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A HYBRID COMPUTER TECHNIQUE FOR SOME OPTIMAL CONTROL PROBLEMS

by



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A THESIS

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ABSTRACT

The theory of optimal control of finite-dimensional differential dynamic systems has been extensively developed. The application of Pontryagin's Minimum Principle often leaves only a two-point boundary-value problem for solution. The solution of these problems is an area of expanding concern as application of optimal control is attempted. Digital computer methods have been developed. In this thesis, the hybrid computer is used in solution of these problems.

A recent hybrid computer technique has been extended to a more general target set case. The underlying assumptions have been shown to be inapplicable in general, thus limiting the class of problems for which the technique is valid. A program using the hybrid computer for applying the technique to second order systems was written and tested.

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MAJOR SYMBOLS

Symbols

C	Cost
D	Difference between target set and state
E	Terminal error
f	State derivative
G	Target set
H	Hamiltonian function
J	Cost functional
k	Positive constant
m	Control space dimension
n	State space dimension
p	Costate
R	Reachable set
t	Time
T	Terminal time
u	Control function
U	Admissible set of controls
x	State

Superscripts

(o)	Initial quantity
$*$	Optimal quantity
$+$	Arbitrary quantity
T	Transpose
\cdot	Time derivative
$-$	Closure of a set
\wedge	Augmented quantity

Subscripts

o	Initial quantity
f	Final quantity
T	Terminal quantity
$0,1,\dots,n$	Coordinate
$-$	Vector or matrix

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CHAPTER I

INTRODUCTION

1.1 The Optimal Control Problem

In recent years, several approaches have been developed for the automatic control of physical systems in an optimal manner. A hybrid computer technique for some optimal control problems is presented in this thesis. The systems to which this technique is applicable fall in the class of finite-dimensional continuous-time differential dynamical systems^{[1]*}. See Appendix 1 for definitions. It is assumed that the systems can be simulated on an analog computer.

The statement of the optimal control problem used in this thesis follows. The plant or process is represented by a vector-matrix differential equation called the state equation

$$\dot{\underline{x}} = \underline{f}(\underline{x}, \underline{u}, t). \quad (1.1)$$

The block diagram of Figure 1.1 represents the system.

The problem is defined over some connected interval of time, I , with

$$t \in I. \quad (1.2)$$

The state, $\underline{x}(t)$, represents the changing characteristics of the system, given by the solution to (1.1). With E^n taken as Euclidean n -space,

$$\underline{x}(t) \in E^n. \quad (1.3)$$

No other constraints are placed on $\underline{x}(t)$.

* Numbers in square brackets refer to the literature cited in the Bibliography.

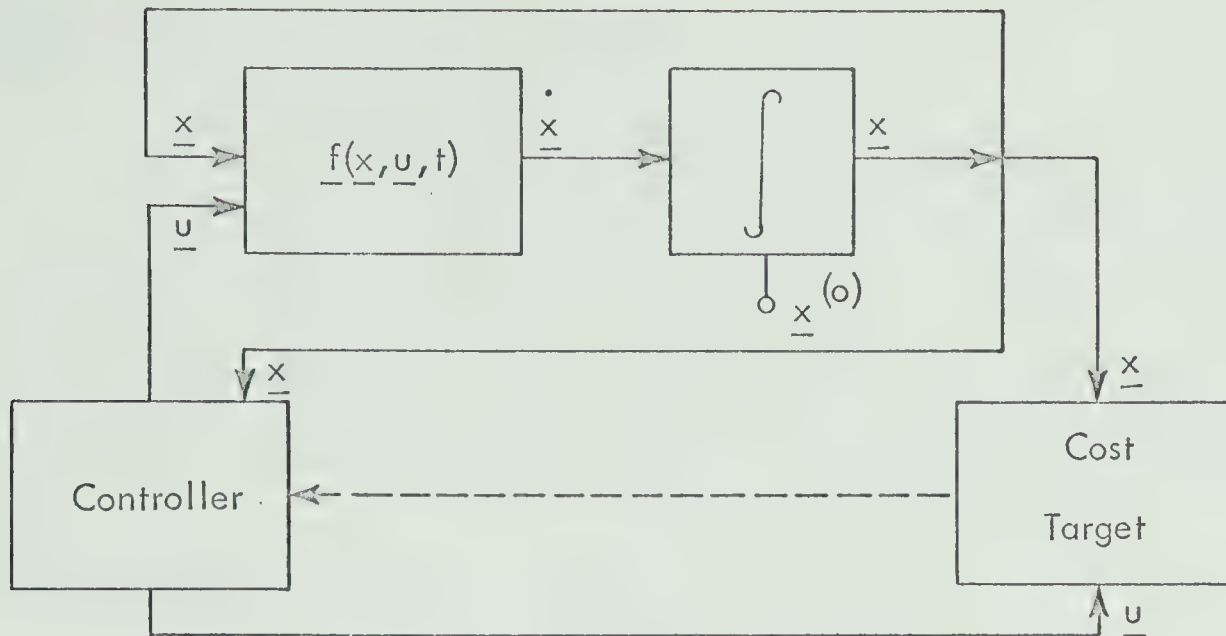


FIGURE 1.1 BLOCK DIAGRAM OF THE SYSTEM

The control function, $\underline{u}(t)$, represents the input to the system by which the output is determined.

$$\underline{u}(t) \in U \subset E^m. \quad (1.4)$$

It is assumed that $\underline{u}(t)$ is not subject to any state-dependent constraints.

The index of performance or cost functional of the system is taken to be the integral of some scalar function of the states, controls and time.

$$J(t) = \int_{t_0}^t f_0(\underline{x}, \underline{u}, \tau) d\tau \quad (1.5)$$

where

$$t \in [t_0, t_f] \subset I. \quad (1.6)$$

The state of the system is initially given by

$$\underline{x}(t_0) = \underline{x}^{(0)}(t_0). \quad (1.7)$$

The final state is required to be

$$\underline{x}(t_f) \in G(\underline{x}, t). \quad (1.8)$$

where

$$G(\underline{x}, t) = \{\underline{x}(t) \in E^n \mid g(\underline{x}, t) \leq 0\}. \quad (1.9)$$

The optimal control problem involves the determination of the optimal control $u^*(t)$ from among the class of admissible controls given by (1.4) which transfers the state $x(t)$ of (1.1) from the initial state (1.7) to the target set (1.8) with the minimum value of cost (1.5).

1.2 Pontryagin's Minimum Principle

The use of Pontryagin's minimum principle is one of several possible approaches to solution of the problem posed in the previous section. The use of this principle yields the necessary conditions which an optimal control function must satisfy. These conditions are also sufficient conditions for all extremal controls. Several excellent textbooks containing a detailed derivation of these conditions are available. Some of these are noted in the bibliography^[1,8,13]. It is assumed in this thesis that the minimum principle can be applied to the systems studied.

The relevant parts of the necessary conditions are briefly stated here.

Define the costate vector, $\underline{p}(t)$, a form of Lagrange multiplier,

$$\underline{p}(t) \in E^n. \quad (1.10)$$

Augment the costate vector with $\underline{p}_0(t)$.

$$\hat{\underline{p}}(t) = (\underline{p}_0(t), \underline{p}(t))^T \in E^{n+1} \quad (1.11)$$

where $(\cdot)^T$ means the transpose.

Augment the state vector with J , by letting

$$\underline{x}_0(t) = J(t); \quad \underline{x}_0(t_0) = 0 \quad (1.12)$$

so that $\dot{\underline{\hat{x}}} = \underline{\hat{f}}(\underline{\hat{x}}, \underline{u}, t) = \underline{\hat{f}}(\underline{x}, \underline{u}, t)$

$$= (f_0(\underline{x}, \underline{u}, t), \underline{f}(\underline{x}, \underline{u}, t))^T. \quad (1.13)$$

Define the Hamiltonian function as

$$H(\underline{\hat{x}}, \underline{u}, t, \underline{\hat{p}}) = \underline{\hat{p}} \cdot \underline{\hat{f}}. \quad (1.14)$$

Clearly

$$\dot{\underline{\hat{x}}}(t) = \frac{\partial H}{\partial \underline{\hat{p}}} . \quad (1.15)$$

To specify $\underline{\hat{p}}(t)$, we define

$$\dot{\underline{\hat{p}}}(t) = - \frac{\partial H}{\partial \underline{\hat{x}}} . \quad (1.16)$$

One of the necessary conditions required by the minimum principle is that the Hamiltonian for an extremal control be minimum with respect to all other controls. That is, for $u^*(t)$ an extremal control

$$H(\underline{\hat{x}}, \underline{u}^*, t, \underline{\hat{p}}) \leq H(\underline{\hat{x}}, \underline{u}, t, \underline{\hat{p}}) . \quad (1.17)$$

This condition will ordinarily determine the extremal controls as a function of the states and costates, found by solving (1.15) and (1.16).

$$\underline{u}(t) = \underline{v}(\underline{\hat{x}}, \underline{\hat{p}}, t) . \quad (1.18)$$

Using these necessary conditions, extremal trajectories can be generated. The problem is reduced to what is termed the two-point boundary-value problem, finding the correct initial conditions for the state and costate which bring the extremal trajectory to the specified final conditions. An extremal trajectory thus found may then be tested for optimality.

1.3 Solution of the Two-Point Boundary-Value Problem

It is clear from the previous section that the solution of the two-point boundary-value problem is essential for the application of the minimum principle. Several methods have been proposed

for solving the two-point boundary-value problem.

For the simplest of problems, analytical solutions may be found. Athans and Falb^[1] develop these techniques for a few systems. In most cases, however, some form of iterative procedure must be used.

One strategy would be to guess a set of unknown initial conditions, solve the system of differential equations for those conditions, and correct the initial conditions so as to better match the final conditions. Thus, this strategy is a parameter search on the missing initial conditions carried out by the iterative solution of a set of initial condition problems. Many other strategies exist, and there are many ways of implementing each. Bekey and Karplus^[3], Balakrishnan and Neustadt^[2], and McLeod^[9] provide starting points for literature surveys. The strategy described here is widely used and will be used in this thesis in a modified form.

Solution may be attempted by pure analog methods, with human decision-making. Using repetitive operation, trial and error searches are feasible on up to third order systems, at least. Some of the articles in McLeod^[9] examine analog methods applied to two-point boundary-value problems. Analog computation is severely disadvantaged by the inability of analog computation to handle anything but the simplest logical decisions.

The usual tool for solving two-point boundary-value problems is the digital computer. One of many possible sources, Balakrishnan and Neustadt^[2] gives a survey of the application of a number of techniques.

Some general concepts of digital computer solution of two-point boundary-value problems using the general strategy described

are presented here.

The accuracy of digital computer solutions can be very high. In cases where the accuracy requirements on the solution are higher than the .1% to 1% possible with modern analog computers, there is no alternative to using a digital computer. Such requirements arise, for example, in satellite trajectory computations. Often, however, the accuracy is a by-product of the necessity for preventing accumulated round-off errors from destroying all accuracy of a digital computer solution, and is not required (or usable) in many engineering applications. Appendix 2 expands these concepts.

In terms of computer utilization, digital computer solutions are often expensive in that they take a large amount of storage and much computation time. Specifically, a fast computer with a large amount of storage is required if a problem is to be solved quickly in real time. Solution on a "small" digital computer will in general be slow in real time and in some cases may be impossible because of lack of storage capability. This situation is caused by two factors in the solution of differential equations by the use of a digital computer. The system must be solved as a set of difference equations, which means that a large number of calculations must be carried out to high accuracy in order to assure an accurate solution and to avoid cumulative round-off and systematic errors which may otherwise occur. Each calculation by a digital computer must be accomplished serially, increasing solution time with system complexity. Use of the analog computer for solution of differential equations does not have these disadvantages.

Hybrid computation has the advantage that the best points of analog and digital computation can be used. The speed of a high speed

analog computer can be used for the solution of the differential equation system. The logical, storage and calculation capabilities of the digital computer can be used for implementation of automatic iteration schemes. The accuracy possible is limited by the analog solution. Further features of hybrid computation are discussed in Appendix 2. In many engineering situations, a fast and inexpensive solution tool is available to do a job which could be done equally well only by a much more expensive digital computer.

1.4 Scope of the Thesis

A computer program for the solution of some optimal control problems is presented. Pontryagin's minimum principle is used to determine a two-point boundary-value problem. A hybrid computer available in the Electrical Engineering Department is used to show that the computational technique is fast compared to some digital computer techniques. Appendix 3 gives a brief description of the hybrid computer used.

The technique presented here was inspired by a paper by T. Miura, J. Tsuda and J. Iwata^[10]. The authors presented their approach for point target sets at the origin. They have given no details of their computational methods.

This thesis presents a derivation of an extension of their work, an improvement in technique of application of the derived conditions, and a limitation of the class of problems for which the technique is applicable.

The derivation is extended to the case of general fixed target sets, with the method of attack for time-varying target sets indicated but not derived. The algorithm derived by Miura et al turns

out to be a special case of the general algorithm.

The technique of application is improved by the use of high speed analog computer techniques to quickly solve the differential equations; by fast digital computation and control by the use of a low-level assembler language; and by the use of a steepest descent search scheme to bring rapid convergence^[5].

The class of problems for which the technique converges to the optimal solution is restricted and it is shown that the derivation in Miura et al^[10] is incorrect as given.

The derivation of the algorithm is given in Chapter 2. The method of solution and the hybrid computer program are discussed in Chapter 3. The application of the technique to two second order systems is discussed in Chapter 4. The fifth and last chapter of the thesis consists of a discussion of the results and suggestions for future work. Four appendices are provided for related reference.

CHAPTER 2

DEVELOPMENT OF THE TECHNIQUE

2.1 Introduction

The technique used to reach the optimal control solution is described in this chapter. This technique is basically a method of solving the two-point boundary-value problem which results when Pontryagin's minimum principle is applied to the control problem. The conditions for convergence of this technique are examined. Both fixed and time-varying target sets are considered. A detailed derivation is included for the case of fixed target sets, whereas no derivation is given for the technique applied to the time-varying case.

2.2 Fixed Target Sets

For each of a monotonically increasing set of costs, the terminal state is brought as close as possible to the target set. This process is terminated when the terminal state is sufficiently close to or at the target set. The value of initial costate which brings this solution is defined as the solution of the two-point boundary-value problem.

Statement of the Problem

Given the dynamical system of (1.13) with

$$\dot{\underline{x}} = \underline{f}(\underline{x}, \underline{u}, t) \quad (2.1)$$

with initial state

$$\underline{x}^{(0)}(t_0) \quad (2.2)$$

and final state in the target set given by (1.9)

$$\underline{x}(t_f) \in G(\underline{x}) \quad (2.3)$$

and controls

$$\underline{u}(t) = \underline{v}(\hat{\underline{x}}, \hat{\underline{p}}) \quad (2.4)$$

from (1.18), find the set of initial conditions, $\underline{\eta}^*$, of the costate equation (1.16)

$$\dot{\hat{\underline{p}}}(t) = - \frac{\partial H}{\partial \hat{\underline{x}}} \quad (2.5)$$

such that (2.1), (2.2), (2.3), (2.4) and (2.5) are all simultaneously satisfied. The solution vector, $\underline{\eta}^*$, is the solution of the two-point boundary-value problem.

It is assumed that there exists a solution $\underline{\eta}^*$ which solves the problem in a finite time, t_f^* , and with a finite minimum cost, C^* .

Before proceeding with the derivation of the technique, a few definitions will be given.

The Banach Space E^n

With the Euclidean norm, $||\underline{x}|| = (\sum_{i=1}^n |x_i|^2)^{1/2}$, $\underline{x} \in E^n$,

and the norm-derived metric $\rho(\underline{x}, \underline{y}) = ||\underline{x} - \underline{y}||$; $\underline{x}, \underline{y} \in E^n$,

ρ being the distance function; the space E^n is a Banach space^[12].

That is, it is a normed (thus metric), linear, and complete space of vectors.

Distance of a vector from a set

The distance from a vector \underline{x} to a set G is defined as

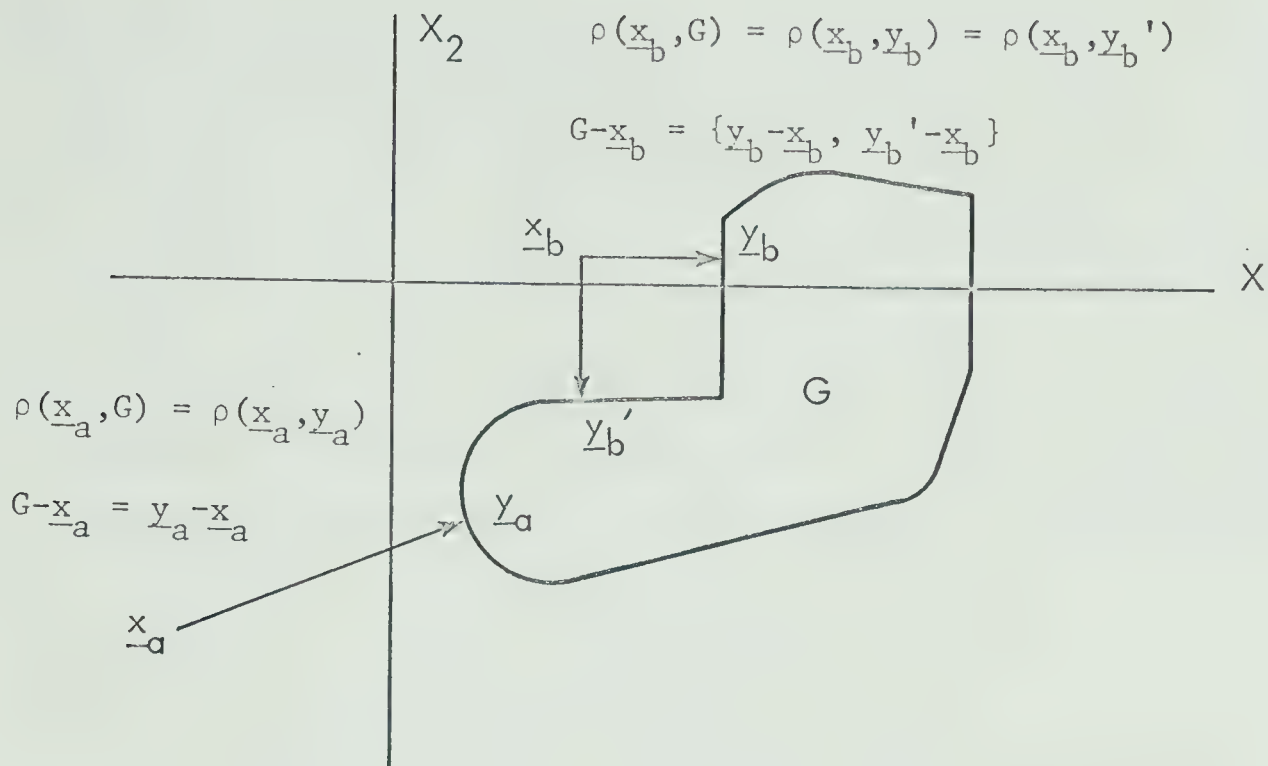
$$\rho(\underline{x}, G) = \inf_{\underline{y} \in G} \rho(\underline{x}, \underline{y}).$$

Vector from a vector to a set

The vectors from a vector \underline{x} to a set G are

$$G - \underline{x} = \{\underline{y} - \underline{x} \mid \rho(\underline{x}, \underline{y}) = \rho(\underline{x}, G), \underline{y} \in \overline{G}\},$$

which may not be unique but which have unique length. These last two concepts are illustrated in Figure 2.1.

FIGURE 2.1 SOME DISTANCE NOTIONS IN E^2

The reachable set, called $R(C)$, is defined as the set of all states which may be reached from $\underline{x}^{(o)}(t_0)$ in cost C using the extremal control strategy. In symbolic notation this can be stated as

$$R(C) = \{ \hat{\underline{x}}(T) \mid \hat{\underline{x}}(T) \in E^{n+1}; J = C \text{ at } t = T; \text{ and } (2.1), (2.2), (2.4), (2.5) \text{ hold} \}. \quad (2.6)$$

Clearly $R(C) \cap G(\hat{\underline{x}}) \neq \phi$ if and only if $C \geq C^*$.

Consider $R(C) \cap G(\hat{\underline{x}}) \neq \phi$, which implies that there exists a vector, $\underline{z}(T)$, such that $\underline{z}(T) \in R(C)$ and $\underline{z}(T) \in G(\hat{\underline{x}})$. But then (2.3) is satisfied as well as the others given in (2.6), the definition of $R(C)$, and the problem is solved. Thus $C = C^*$ since this is the minimum extremal solution. The reverse case is obvious. Thus $R(C) \cap G(\hat{\underline{x}})$ is nonempty first at $C = C^*$. The approach used is based on this observation.

Derivation of the Method

The underlying principle of the method is outlined here.

Beginning with a cost of zero and using an increasing set of costs, the terminal state (in the reachable set) is kept as close as possible to the target set. If this condition is maintained from $C = 0$ until $C = C^*$, then the initial costate corresponding to the last iteration which had $C = C^*$ and which brought the terminal state to the target set is the one which produces the optimal control, because $R(C^*) \cap G(\hat{\underline{x}}) = \hat{\underline{x}}(t_f)$ and equations (2.1) to (2.5) are all satisfied.

Formulation of an Equivalent Problem

Consider operating the system* with a particular cost C and an initial costate $\underline{\eta}$. The terminal states and costates can be considered to be functions of the parameter C and the vector parameter $\underline{\eta}$, since fixing these values and operating the system must give a unique result. The terminal state and costate are thus written as $\hat{\underline{x}}_T(C, \underline{\eta})$ and $\hat{\underline{p}}_T(C, \underline{\eta})$. Since $G(\hat{\underline{x}})$ is a set in E^{n+1} independent of C and $\underline{\eta}$, G can be written for $G(\hat{\underline{x}})$. Then the vector $G(\hat{\underline{x}}) - \underline{x}(C, \underline{\eta})$ can be written $G - \underline{x}(C, \underline{\eta}) = \underline{D}(C, \underline{\eta}) \in E^{n+1}$. Of course,

$$||\underline{D}(C, \underline{\eta})|| = ||G - \hat{\underline{x}}(C, \underline{\eta})|| > 0 \quad \text{for } C < C^*.$$

The problem to be solved is that of finding the correct value of initial costate for any cost C . Consider any value of $\underline{\eta}$, say $\underline{\eta}^+$. At t^+ , $J = C$ and $\underline{D}^+ = \underline{D}(C, \underline{\eta}^+)$. \underline{D}^+ is a fixed vector in E^{n+1} . The new objective is to minimize the cost functional

$$J' = k(\underline{D}^+, \underline{D}(C, \underline{\eta})) \quad (2.7)$$

* The phrase "operating the system" is taken to mean the following. For a given problem, assign a value to $\underline{\eta}$, the initial costate and solve the system equations until the cost is C . The time at which this occurs is called the terminal time, T . The state and costate are the terminal state and costate, $\hat{\underline{x}}(T)$ and $\hat{\underline{p}}(T)$, respectively.

where the inner product used is $(\underline{a}, \underline{b}) = \underline{a}^T \underline{b}$, $\underline{a}, \underline{b} \in E^n$.

There is at least one $\underline{y} \in G$ such that

$$G - \hat{\underline{x}}(C, \underline{\eta}) = \underline{y} - \hat{\underline{x}}(C, \underline{\eta}) = \underline{D}(C, \underline{\eta}). \quad (2.8)$$

These values of \underline{y} (if not unique) are functions of the state $\hat{\underline{x}}$, and all give rise to the same value of $||\underline{D}||$.

Constraints

There are two possible characteristics of G for any \underline{y} . The tangent plane may be unique, in which case $\underline{D}(C, \underline{\eta})$ is normal to all vectors in this tangent plane. The tangent plane to G may not exist at \underline{y} , in which case, if one considers a hypersphere around \underline{y} , $\underline{D}(C, \underline{\eta})$ can be considered as normal to the hypersphere for any non-zero radius. It will be a computational necessity to have a small tolerance around such a point, thus the use of a tangent plane to a hypersphere around the point is a useful concept. Figure 2.2 illustrates these concepts in E^3 .

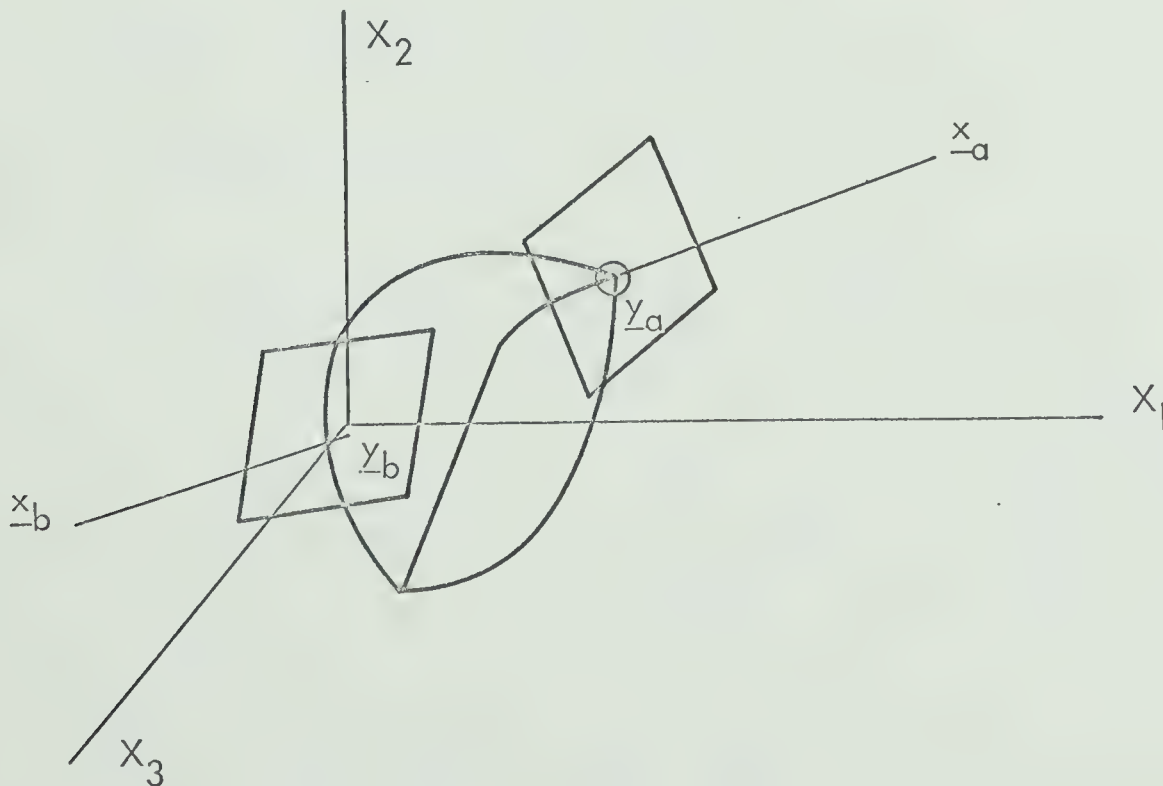


FIGURE 2.2 TANGENT HYPERPLANES IN E^3

Restrictions are placed on the state space and the target set.

In general the cost is unconstrained so that

$$G(\hat{\underline{x}}) = G(\underline{x}) . \quad (2.9)$$

It is assumed that the space E^n is separated into two sets G and $H = G^c (= \{\underline{x} | \underline{x} \in E^n \text{ and } \underline{x} \notin G\})$. $H = \bigcup_{\alpha \in A} H_\alpha$, with each H_α being characterized by a specific value of \underline{x}_α and a constant k_α , and with A a set of indices, quite possibly nondenumerable. For $\underline{x} \in H_\alpha$,

$$\underline{D} = \frac{(\underline{x} - \underline{x}_\alpha)}{\|\underline{x} - \underline{x}_\alpha\|} (|\|\underline{x} - \underline{x}_\alpha\|| - k_\alpha) , \quad \underline{x} \in H_\alpha . \quad (2.10)$$

When $\underline{x} \in H_{\alpha 1} \cap H_{\alpha 2}$ (2.11)

$$\begin{aligned} \underline{D} &= \frac{(\underline{x} - \underline{x}_{\alpha 1})}{\|\underline{x} - \underline{x}_{\alpha 1}\|} (|\|\underline{x} - \underline{x}_{\alpha 1}\|| - k_{\alpha 1}) \\ &= \frac{(\underline{x} - \underline{x}_{\alpha 2})}{\|\underline{x} - \underline{x}_{\alpha 2}\|} (|\|\underline{x} - \underline{x}_{\alpha 2}\|| - k_{\alpha 2}) . \end{aligned} \quad (2.12)$$

This form of \underline{D} means that the following equality holds.

$$\frac{\underline{D}}{\|\underline{D}\|} = \frac{G - \underline{x}}{\|G - \underline{x}\|} = \frac{\underline{x}_\alpha - H_\alpha}{\|\underline{x}_\alpha - H_\alpha\|} , \quad \forall \underline{x} \in H_\alpha . \quad (2.13)$$

\underline{x}_α can be considered to be the "attracting point" for \underline{D} corresponding to each \underline{x} .

In order to analyse the system on a computer, the sets H_α must be grouped into a finite collection H_β , $\beta \in B$, with $\bigcup_{\beta \in B} H_\beta = \bigcup_{\alpha \in A} H_\alpha = G^c$. The unifying feature of each H_β is the function which determines the set of attracting points \underline{x}_α for all $H_\alpha \in H_\beta$. For example, if a set of \underline{x}_α form a line segment, then H_β would be the domain of state space using that line segment as a set of attracting points, and the line segment (and thus each \underline{x}_α) would be given by a particular function

of \underline{x} , for $\underline{x} \in H_\beta$. These concepts are demonstrated in the following examples and in Figures 2.3 to 2.6.

Examples

1. Hyperspheres - Figure 2.4

$$G = \{\underline{g} \mid ||\underline{g} - \underline{g}_0|| \leq R ; \underline{g}, \underline{g}_0 \in E^n\} \quad (2.14)$$

$$\underline{D} = \frac{(\underline{g}_0 - \underline{x})}{(||\underline{g}_0 - \underline{x}||)} (||\underline{g}_0 - \underline{x}|| - R) \quad \underline{x} \in G^c. \quad (2.15)$$

In the framework of the derivation, only one region of $H = G^c$ need be considered, with

$$\underline{x}_\alpha = \underline{x}_\beta = \underline{g}_0, k_\alpha = k_\beta = R \quad (2.16)$$

for a single α and β , $\underline{x} \in H_\alpha = H_\beta \quad \forall \underline{x} \in G^c$.

The work of Miura, et al^[10] is concerned with a special case of this example. The technique is derived there for the target set being the origin of state space. In their paper, then, they have used $\underline{g}_0 = \underline{0}$ and $R = 0$, so that $\underline{D} = -\underline{x}$. Their result is

$$\underline{p}(\tau) = -k\underline{x}(\tau) \quad (2.17)$$

which is the same as given by (2.35) in this case,

$$\underline{p}_T(C, \underline{n}) = -k\underline{x}(C, \underline{n}). \quad (2.18)$$

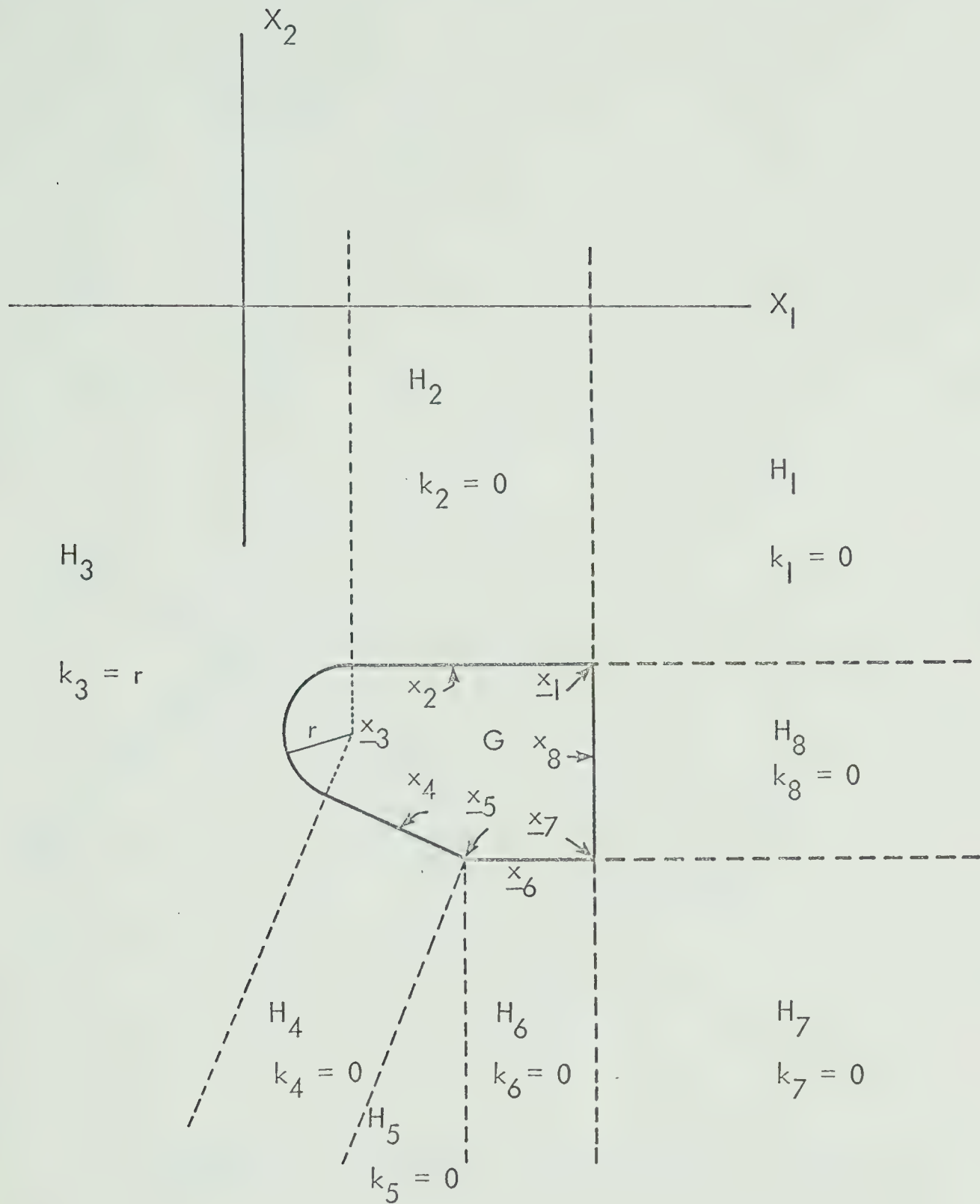
2. Infinite Hypercylinders - Figure 2.5

$$\begin{aligned} g_\ell = \{\underline{g} \mid \underline{g} = \alpha \underline{g}_1 + (1 - \alpha) \underline{g}_2 ; \underline{g}_1, \underline{g}_2, \underline{g} \in E^n; \\ \alpha \in (-\infty, \infty)\}. \end{aligned} \quad (2.19)$$

a line in state space through \underline{g}_1 and \underline{g}_2 .

$$G = \{\underline{g} \mid ||\underline{g} - \underline{g}_\ell|| \leq R; \underline{g} \in E^n\}. \quad (2.20)$$

Here $H_\alpha = \{\underline{x} \mid (\underline{x} - \underline{g}_\ell) = (\underline{x} - \underline{g}_{\ell\alpha}); \underline{g}_{\ell\alpha} \in g_\ell \text{ for } \alpha \in (-\infty, \infty); \underline{x} \in H\}.$ (2.21)



$\underline{x}_1, \underline{x}_3, \underline{x}_5, \underline{x}_7$ are points

$\underline{x}_2, \underline{x}_4, \underline{x}_6, \underline{x}_8$ are line segments

FIGURE 2.3 ATTRACTION FUNCTIONS IN STATE SPACE

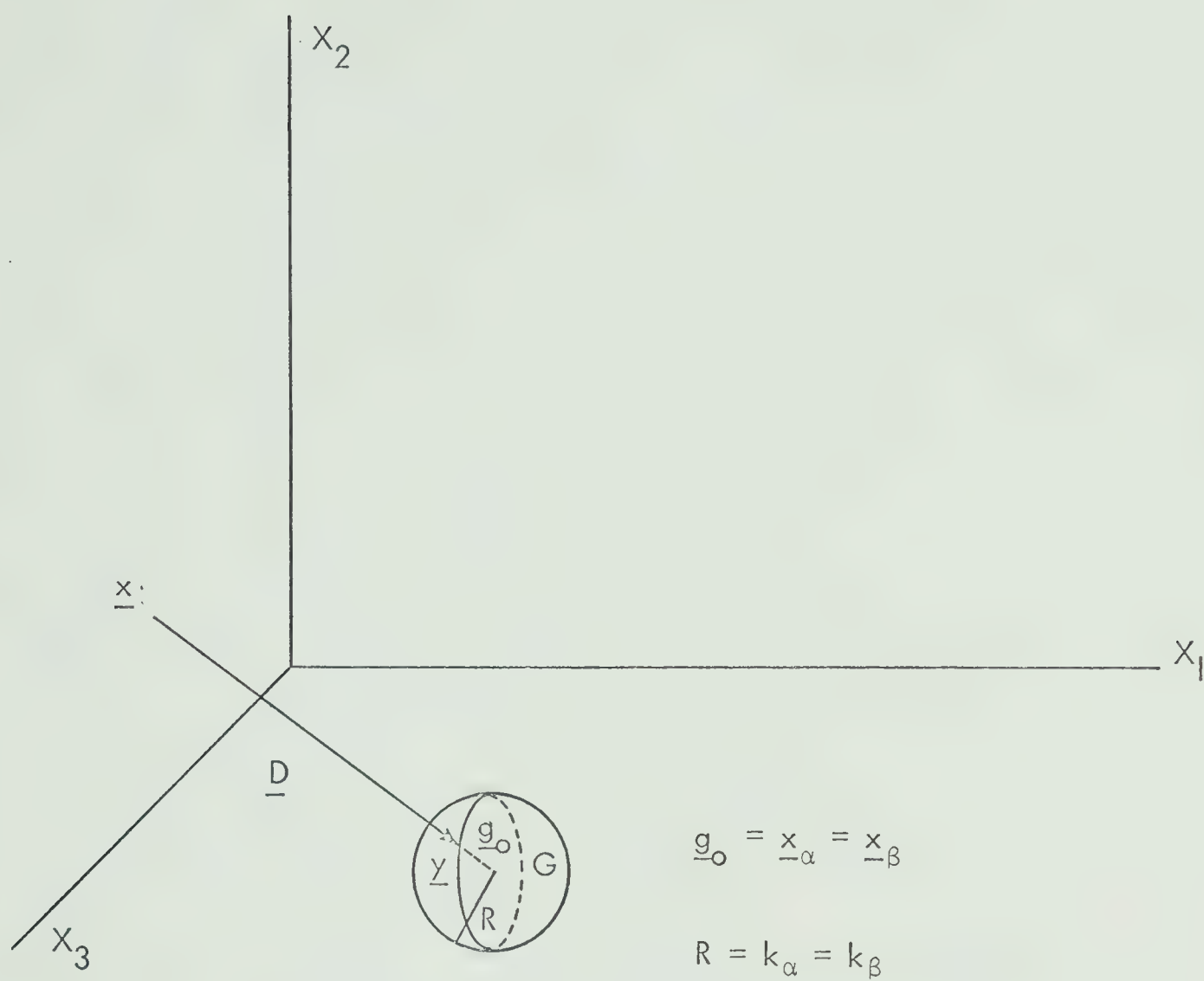


FIGURE 2.4 EXAMPLE I: HYPERSPHERICAL TARGET SET

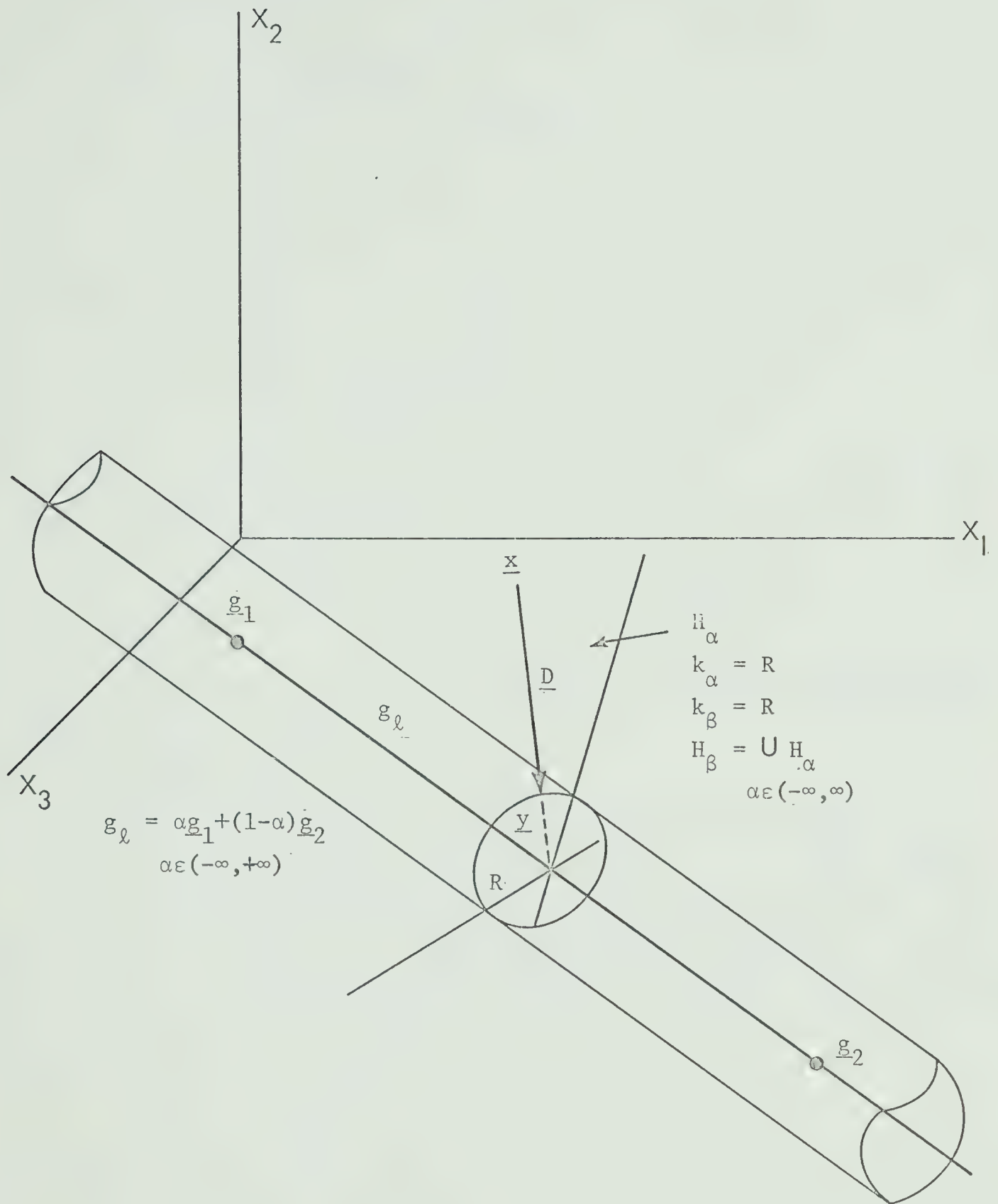


FIGURE 2.5a EXAMPLE 2: INFINITE HYPERCYLINDRICAL TARGET SET

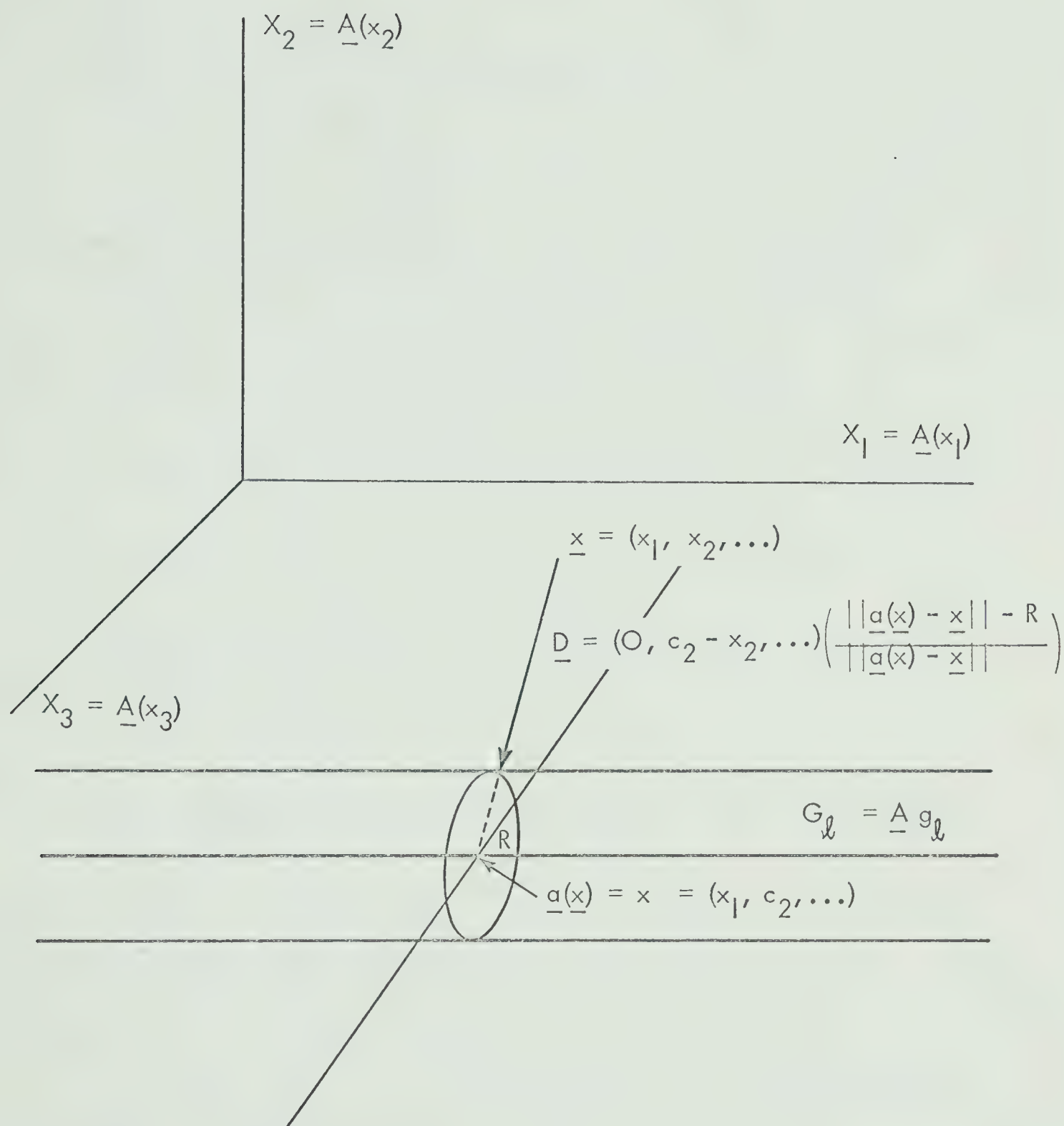


FIGURE 2.5b EXAMPLE 2: INFINITE HYPERCYLINDRICAL TARGET SET IN TRANSFORMED STATE SPACE

$$\underline{x}_\alpha = \underline{g}_{\ell\alpha} = \alpha \underline{g}_1 + (1 - \alpha) \underline{g}_2 \quad k_\alpha = R$$

$$H_\beta = \bigcup H_\alpha \quad \alpha \in (-\infty, \infty)$$

\underline{x}_β is the set of points forming \underline{g}_ℓ

$$k_\beta = R \quad \text{for a single } \beta \quad (2.22)$$

Here again only one region need be considered with \underline{g}_ℓ being the attracting function (representing a set of attracting points).

Computational considerations indicate that an effective method for treating this type of target set would be by transformation of the system equations by the linear transformation matrix \underline{A} with unity norm which takes the set \underline{g}_ℓ into a line in (say) the x_1 direction.

$$\underline{A}\underline{g}_\ell = \underline{G}_\ell = \{\underline{x} | \underline{x} = (\alpha, c_2, c_3, \dots, c_n); \alpha \in E^1; c_2 \text{ to } c_n \text{ are real constants}\}. \quad (2.23)$$

Assuming that the system equations have been so transformed, we have $\underline{a}(\underline{x}) = (x_1, c_2, c_3, \dots, c_n)$ (2.24)

and

$$\underline{D} = \frac{(\underline{a}(\underline{x}) - \underline{x})}{\|\underline{a}(\underline{x}) - \underline{x}\|} \quad (\|\underline{a}(\underline{x}) - \underline{x}\| = R). \quad (2.25)$$

Note that $\underline{a}(\underline{x}) - \underline{x} = (0, c_2 - x_2, c_3 - x_3, \dots, c_n - x_n)$. (2.26)

This makes the calculation of \underline{D} convenient in practice.

3. Hypercylinders with Hyperhemispherical Ends - Figure 2.6

$$\underline{g}_\ell = \{\underline{g} | \underline{g} = \alpha \underline{g}_1 + (1 - \alpha) \underline{g}_2; \underline{g}_1, \underline{g}_2, \underline{g} \in E^n; \alpha \in [0, 1]\} \quad (2.27)$$

a line segment in state space from \underline{g}_1 to \underline{g}_2 .

$$\underline{G} = \{\underline{g} | \|\underline{g} - \underline{g}_\ell\| \leq R; \underline{g} \in E^n\}. \quad (2.28)$$

Consider a transformation \underline{A} as in the previous example. Then

$$\underline{b}_1 = \underline{A} \underline{g}_1 = (\alpha_1, c_2, c_3, \dots, c_n) \quad (2.29)$$

$$\underline{b}_2 = \underline{A} \underline{g}_2 = (\alpha_2, c_2, c_3, \dots, c_n)$$

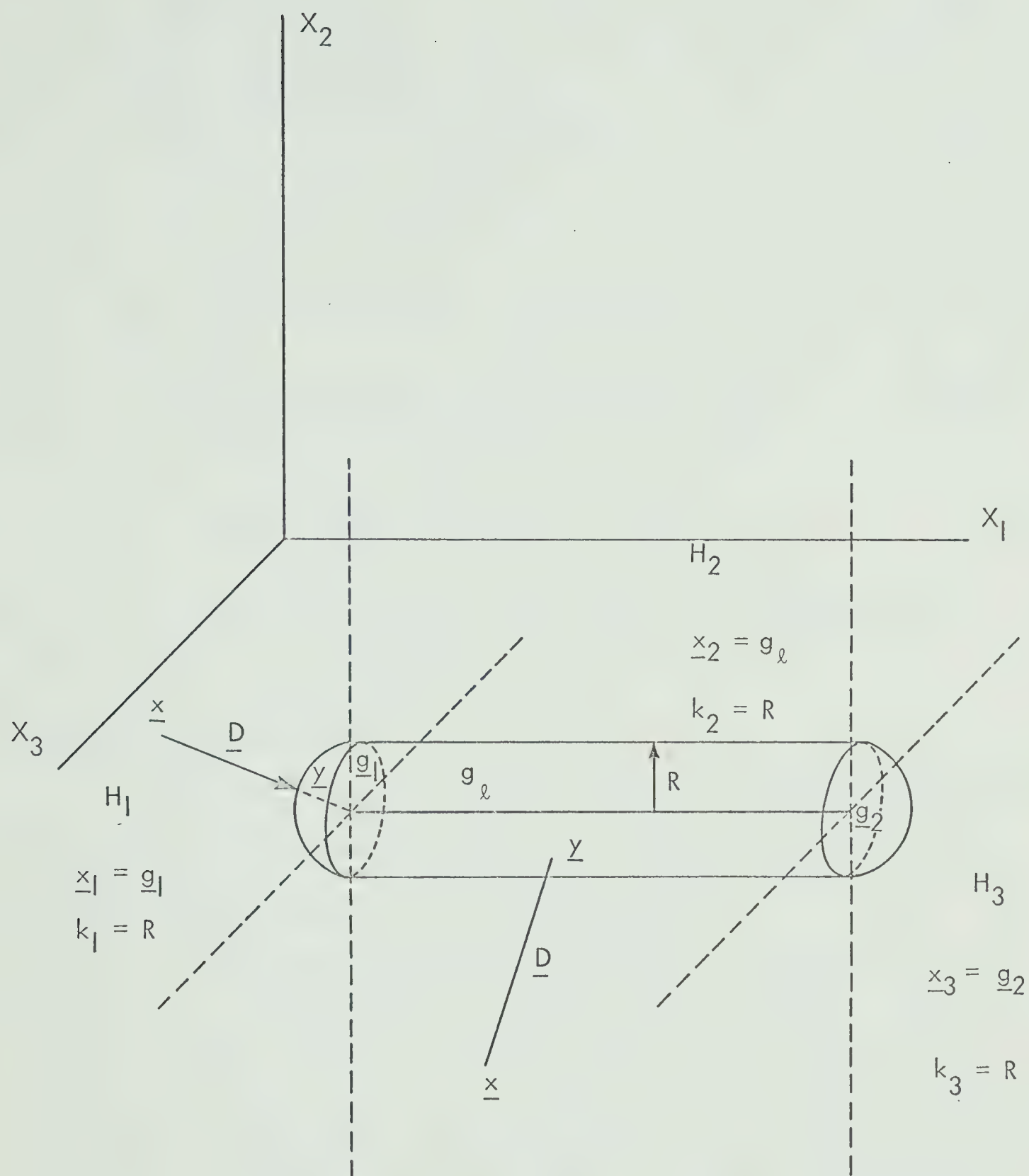


FIGURE 2.6 EXAMPLE 3: HYPERCYLINDRICAL TARGET SET WITH HYPERHEMISPHERICAL ENDS

$$\underline{a}(\underline{x}) = (x_1, c_2, c_3, \dots, c_n) \quad (2.30)$$

$$H_{\beta 1} = \{\underline{x} | x_1 \leq \alpha_1, \underline{x} \in G^c\}$$

$$H_{\beta 2} = \{\underline{x} | \alpha_1 \leq x_1 \leq \alpha_2, \underline{x} \in G^c\} \quad (2.31)$$

$$H_{\beta 3} = \{\underline{x} | \alpha_2 \leq x_1, \underline{x} \in G^c\}$$

$$\underline{D} = \begin{cases} \frac{(\underline{b}_1 - \underline{x})}{\|\underline{b}_1 - \underline{x}\|} & (|\|\underline{b}_1 - \underline{x}\|| - R) \quad \underline{x} \in H_{\beta 1} \\ \frac{(\underline{a}(\underline{x}) - \underline{x})}{\|\underline{a}(\underline{x}) - \underline{x}\|} & (|\|\underline{a}(\underline{x}) - \underline{x}\|| - R) \quad \underline{x} \in H_{\beta 2} \\ \frac{(\underline{b}_2 - \underline{x})}{\|\underline{b}_2 - \underline{x}\|} & (|\|\underline{b}_2 - \underline{x}\|| - R) \quad \underline{x} \in H_{\beta 3} \end{cases} \quad (2.32)$$

In this case there are three regions of state space with different attracting functions.

The Terminal Condition

The purpose of the preceding discussion has been to develop arguments for the terminal condition, the requirement which ensures that the cost functional J' of (2.7) is minimized. It is required that $\underline{D}(C, \underline{\eta})$ be brought as far as possible in the direction \underline{D}^+ . One of the necessary conditions is the transversality condition, that the costate vector be normal to the tangent plane to the target set. The consideration of an infinitesimal hypersphere around a point at which the tangent plane does not exist extends the transversality condition to such points, within tolerances imposed by computational considerations.

Suppose \underline{D}^* is chosen to make J' minimum. Then, for all possible \underline{D} , $k(\underline{D}^*, \underline{D}^*) \leq k(\underline{D}^*, \underline{D})$ (2.33)

$$\text{or} \quad ||\underline{D}^*|| \leq ||\underline{D}||. \quad (2.34)$$

Thus \underline{D}^* is the vector to the target set resulting from the terminal state $\underline{x}_T^*(C)$ closest to the target set.

The requirement for finding \underline{D}^* is that J' be minimized.

From the transversality condition a necessary condition for this is

$$p_T(C, \underline{\eta}) = k \underline{D}(C, \underline{\eta}) \quad (2.35)$$

since $\underline{D}(C, \underline{\eta})$ is normal to the tangent plane to the target set or to the infinitesimal hypersphere around appropriate points of the target set. (2.35) is called the terminal condition in this thesis.

Application of the Terminal Condition

The extremal control generated by the terminal condition is found in the following manner. At cost $C = 0$, the value for $\underline{\eta} = k\underline{D}$ is assigned, which satisfies (2.35) since $C = 0$ implies $t = 0$. A new cost is formed by incrementing the cost by ΔC . Beginning with the value of $\underline{\eta}$ from the previous step, a new value is found which satisfies (2.35). This process is repeated until the terminal state is close enough to the target set. In this manner, the terminal state at each cost C has been brought as close as possible to the target set. For the case where the reachable set is convex and for the case where the local minimum of J' which first appears eventually leads to the optimal solution, this strategy will find the optimal solution. These considerations will be discussed later in this chapter in section 2.4.

The search strategy by which the correct value of $\underline{\eta}$ is found at each cost C consists of a steepest descent search applied to the minimization of a terminal error norm.

The error norm used is of the following form. The terminal condition required that (2.35) is satisfied. This will be so if and

only if

$$\frac{p_T(C, \underline{n})}{||p_T(C, \underline{n})||} = \frac{D(C, \underline{n})}{||D(C, \underline{n})||} \quad (2.36)$$

Thus, the vector $\underline{E}(C, \underline{n})$ defined by

$$\underline{E}(C, \underline{n}) = \frac{p_T(C, \underline{n})}{||p_T(C, \underline{n})||} - \frac{D(C, \underline{n})}{||D(C, \underline{n})||} \quad (2.37)$$

represents the error in direction of p_T . The norm of $\underline{E}(C, \underline{n})$ is used as the terminal error norm.

The search strategy consists of five steps. Two tests are made at various steps. Each time the system is operated, $||\underline{D}||$ is tested. If it is less than the error tolerance the search for \underline{n} is ended. Each time $||\underline{E}||$ is calculated, it is tested. If it is less than the error tolerance, the terminal condition is considered met and the cost C is increased by the increment ΔC . The search goes back to step 1.

1. The previous successful value of \underline{n} is used as a starting value and the system is operated. For the first step, $\underline{n} = k\underline{D}$ is used.
2. The system is operated n times with steps $\Delta \eta$ in each of the (Cartesian) coordinates of \underline{n} .
3. From these $n + 1$ values of the terminal error norm, the negative of the gradient of the terminal error norm, $-\bar{\nabla}||\underline{E}||$, is found, as a first order approximation. This is the direction of steepest descent of $||\underline{E}||$.
4. The value \underline{n} is modified by the addition of the term $(-\bar{\nabla}||\underline{E}||)\Delta \eta$ and the system is operated.
5. Step 4 is repeated until $||\underline{E}||$ no longer improves. When this occurs, step 4 is reversed by subtraction of the correction term

to give the last value of \underline{n} which improved $||\underline{E}||$. Then the search returns to step 1.

The flow chart in Figure 2.7 shows the search logic.

2.3 Time-Varying Target Sets

The natural extension of the preceding section is to a time-varying target set, using the terminal time T each time the system is operated as the time for defining the target set in finding a value for \underline{D} . That is, if the target set is a time-varying set, $G(t)$, then when the system has been operated

$$\underline{D}(C, \underline{n}) = G(T) - \underline{x}(C, \underline{n}) \quad (2.38)$$

Used in this manner, the correct final time is an additional parameter dependent on determination of the extremal solution. The same arguments hold as in the previous section. In particular, at the closest approach of $R(C)$ to the target set, there would exist a $\underline{D}(C, \underline{n})$ which minimizes J' from (2.7). The application of (2.35) would find the value of \underline{D} for each C .

A second approach is possible. The final time is predicted and called T_f and $G(T_f)$ is used as the target set. Then T_f is modified in the iterations along with C or \underline{n} . \underline{D} could be treated as a function of T_f , so that a new value of T_f would be found using $||\underline{D}|| = 0$ and an interpolation scheme. Substantial programming effort would probably be necessary to ensure an orderly automatic solution attempt.

2.4 Convergence

There are two important considerations in the convergence of the technique presented. The conditions affecting the convergence to an extremal solution are discussed, as well as the conditions under which the solution found is optimal.

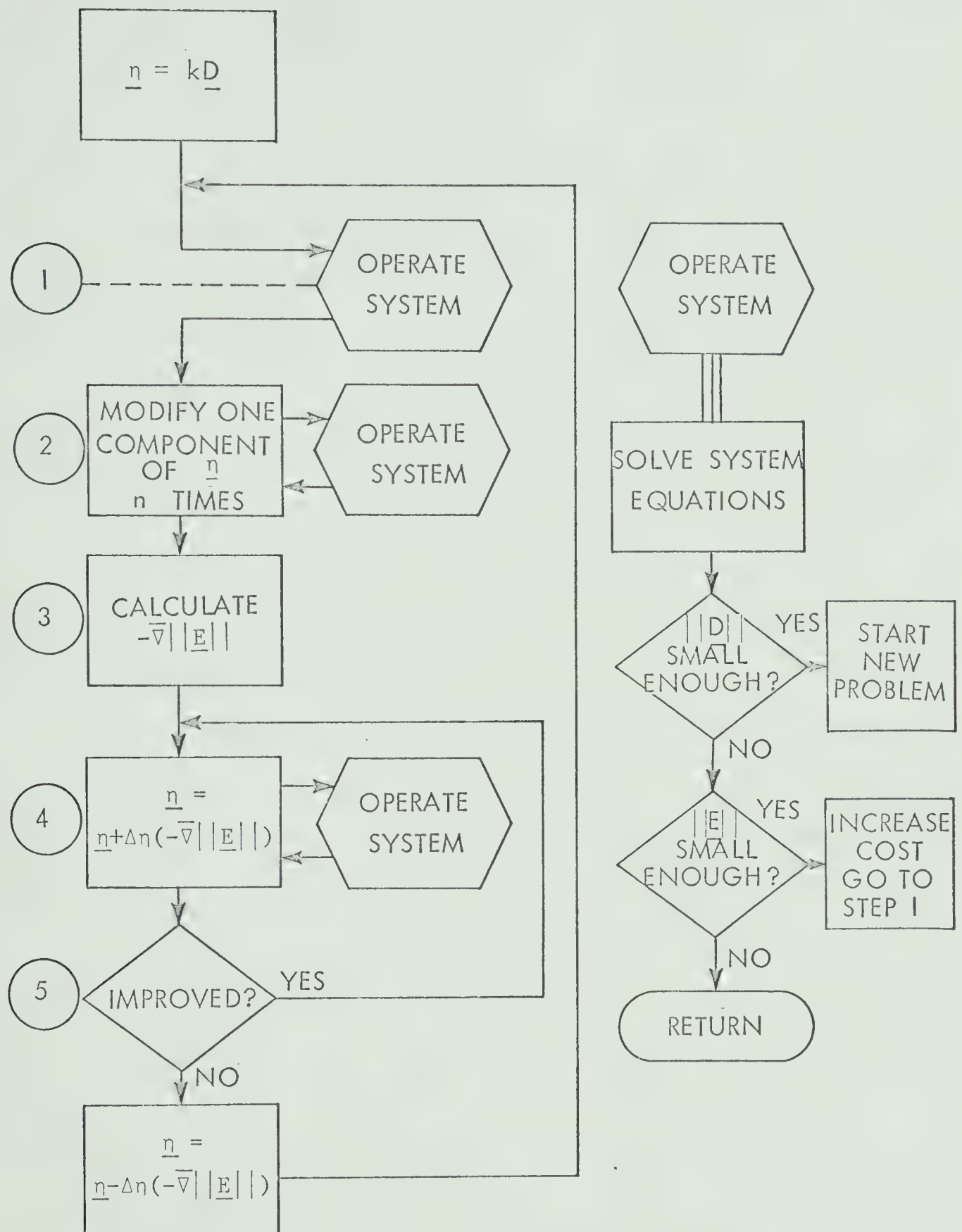


FIGURE 2.7 THE SEARCH LOGIC

It is not possible to exactly specify the requirements which would guarantee convergence to an extremal solution. As in all discrete-step methods of solving continuous problems, step size of the various incremented quantities can be of fundamental importance. The sensitivity of the final state to the initial costate often becomes great, requiring larger error tolerances in terminal matching and final state conditions. Qualitatively, it is possible to say that this method guarantees convergence in the limit as step sizes become small enough and if the costates are continuous functions of time. The latter condition, along with the continuity of the cost guarantees a continuous variation of initial costate with respect to cost. This ensures that the new initial costate will be close to the old one, for small enough changes in the cost. This heuristic argument is intended only as an explanation for the convergence of the method which has been observed.

In order to facilitate convergence the following procedure is used. Initially, rather large incremental steps are used for the search. When the terminal state is brought to within the initial error specified for the target set, the size of the steps used is reduced. The solution is then continued from that point with smaller steps, allowing close approach of the terminal state to the target set. This approach, often called "getting into the ball park", has been found to give rapid convergence in real time as applied.

This rapid convergence feature gives this method an advantage over pure analog and pure digital solutions. Pure analog solutions cannot be automatically implemented. Pure digital methods, following the strategy of a parameter search on the missing initial conditions as introduced previously, often require initial trial values close to the final solution in order to guarantee convergence without an

inordinate number of iterations, which are costly in real time and capacity. The method proposed here uses the ease of solving differential equations to advantage in order to ensure convergence to an extremal solution.

The conditions under which the extremal solution found by this method is the optimal solution have been investigated. Consider a convex target set. When the reachable set $R(C)$ is convex, convergence is guaranteed to the optimal solution. This condition is not usually satisfied. When the reachable set is not convex, there may appear, for costs greater than some cost C_f , more than one point in $R(C)$ $C > C_f$ which are local minima of J' . That is, there may be multiple points for which J' is locally minimum. The method proposed here will find one of them, usually the one for which $\left. \frac{d||D||}{dC} \right|_{C_f}$ is most negative. This method searches for the extremal solution which most rapidly approaches the target set at the cost where two or more local minima of J' come into existence in $R(C)$. This point will be further clarified by means of an example.

Example

Consider the system given by the following specifications.

$$\dot{\underline{x}} = \begin{bmatrix} 1 \\ u \end{bmatrix} \quad \underline{x}^{(0)} = \begin{bmatrix} -10 \\ 0 \end{bmatrix} \quad \underline{x}^{(f)} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad (2.39)$$

$$u \in [-1, 1] \quad (2.40)$$

$$J = \int_0^{t_f} C(x_2) dt \quad (2.41)$$

$$C(x_2) = \begin{cases} .1 - (x_2 + 1.5) & x_2 \leq -1.5 \\ 10 & -1.5 < x_2 < 1.0 \\ 5 + (x_2 - 1) & 1.0 \leq x_2 \end{cases} \quad (2.42)$$

Examination of the system shows that the optimal solution strategy consists of the control sequence $\{-1; 0; 1\}$ switching at times $\{0; 1.5; 8.5\}$ and with cost 30.7. A second extremal trajectory is caused by the control sequence $\{1; 0; -1\}$ switching at times $\{0; 1; 9\}$ with cost 60. All trajectories remaining in the region $-1.5 < x_2 < 1.0$ are extremal with cost 100. The method presented here would, under near-perfect convergence, maintain the initial value of \underline{n} chosen until $C=100$, since the initial value of \underline{n} satisfies the terminal condition for all costs from 0 to 100. If the program, by drift or noise, should find the branch of $R(C)$ at $x_2 = 1.0$ after cost $C=10.0$, it would attempt to continue with that branch until $C=60$. This would then be indicated as the solution. A rather advantageous set of errors in the program would be required to find the optimal branch of the reachable set. See Figure 2.8.

This example shows that the method may give quite misleading results. Further work would be desirable in clarifying the conditions causing failure of the method to find the optimal solution.

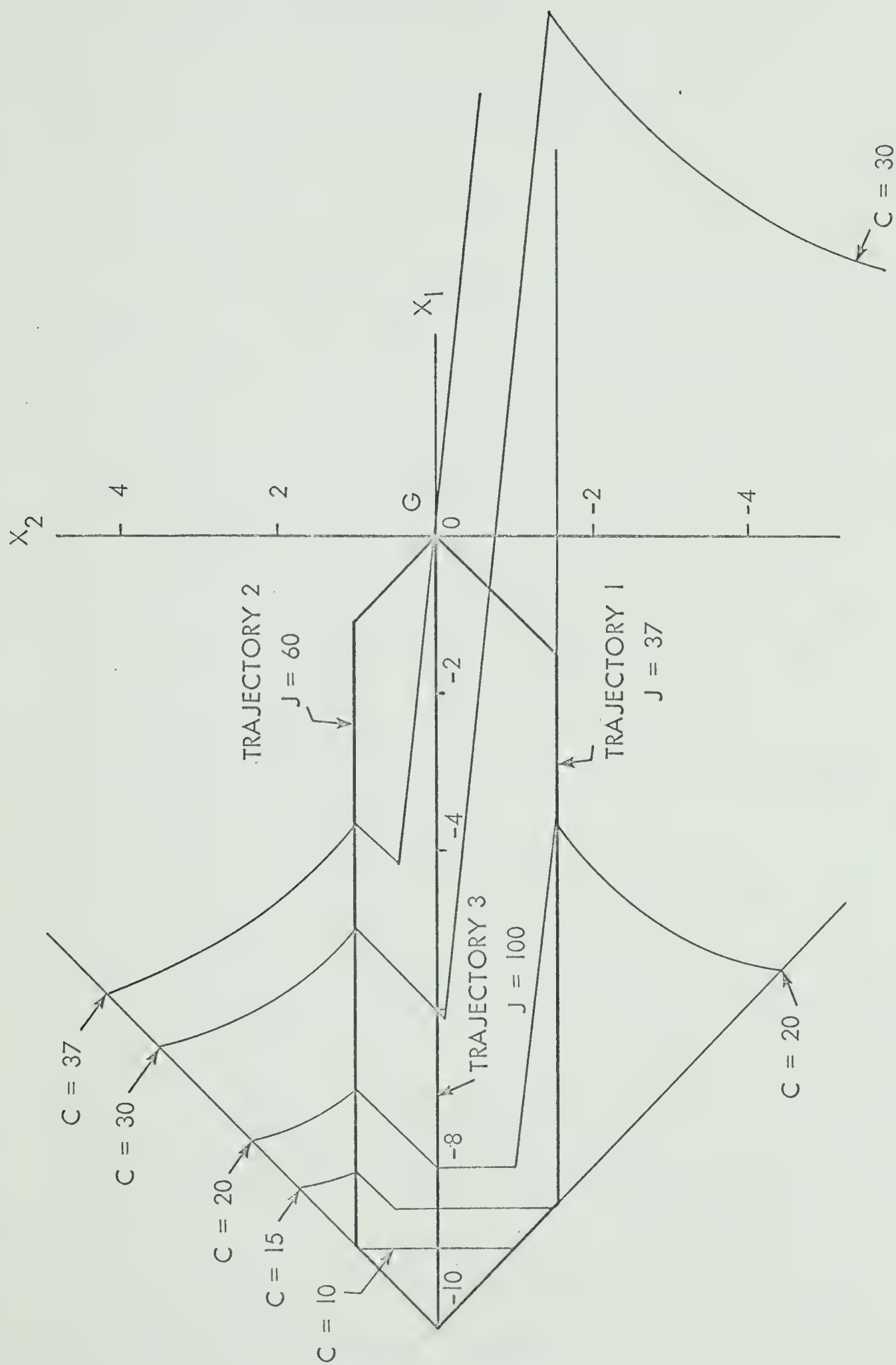


FIGURE 2.8 REACHABLE SETS FOR EXAMPLE
(RIGHT MOST BOUNDARIES SHOWN)

CHAPTER 3

METHOD OF SOLUTION

3.1 Introduction

In this chapter, the technique developed in Chapter 2 is applied to second order optimal control problems. The logic and organization of the program is described. Performance, improvements and extensions of the program are examined. In connection with this chapter, the last three appendices contain supplementary information. The techniques of hybrid computation as applied to this thesis are discussed in Appendix 2. Appendix 3 describes the hybrid computer used, an EAI TR-48 with a DEC PDP-8, in the Electrical Engineering Department at the University of Alberta. The program flow charts and the program listing assembled by PAL-D form the fourth appendix.

3.2 Organization of the Program

Implementation of the Solution Technique

Use of the technique developed in solution of the two-point boundary-value problem involves a basic change in search strategy from the direct parameter search. In the parameter search for the satisfactory unknown initial conditions, the extremal control strategy is implemented while reducing an error norm, the distance from the terminal state to the target set. The problem can be highly complicated by the lack of a final time, which causes complications in the detection of run-end conditions. In the method presented here, the magnitude of the search problem is reduced, and run-end conditions are very simple to handle. The minimization of terminal error at the projected final

time is replaced by the minimization of error in satisfying the terminal direction conditions on the terminal state and costate. At the conclusion of the search, an extremal solution to the optimal control problem is found.

The terminal direction condition, (2.35), is equivalent to the following set of conditions in the second order case.

$$\frac{p_1}{||p_1||} = - \frac{D_1}{||D_1||} \quad \text{or} \quad p_1' = - D_1' \quad (3.1)$$

and

$$\frac{p_2}{||p_2||} = - \frac{D_2}{||D_2||} \quad \text{or} \quad p_2' = - D_2' \quad (3.2)$$

where p_i' and D_i' indicate the normalized quantities as shown. The terminal error norm used is that of (2.37)

$$ND = ((p_1' + D_1')^2 + (p_2' + D_2')^2)^{\frac{1}{2}}. \quad (3.3)$$

The program attempts to bring ND to less than ER, the initially specified terminal error tolerance.

The search scheme devised is a type of steepest descent technique using the gradient of ND at a nominal initial costate. The nominal costate vector is successively perturbed in each component and the approximate gradient calculated from the corresponding values of ND. Then steps are made in the nominal costate vector in the negative gradient direction until no further improvement in ND occurs. The calculation of the gradient is then repeated and further steps in the initial costate are taken. The search ends when the value of ND at any step is less than the terminal error tolerance, ER. Then the cost is increased and the search for the correct initial costate is repeated.

The search for the extremal solution to the original problem is thus occurring as the cost is increased. It was established in Chapter 2 that continuously satisfying the terminal condition would bring the terminal state to the target set when the cost reached the minimum cost or some other extremal cost. When the state error norm

$$NX = (D_1^2 + D_2^2)^{\frac{1}{2}} \quad (3.4)$$

is found to be less than EX, the state error tolerance, the problem is considered to be solved and the solution is printed out.

Task Assignment in the Program [3],[4],[10]

Use of a hybrid computer allows a choice in the manner of execution of the various component tasks which make up the solution of the problem. In this section, brief explanation of the task assignment is given.

Input of the problem data is done by means of the digital computer, programmed to read the teletype keyboard. This allows convenient input to the program with a printed record made.

Using input data or iteration search logic and calculations, the digital computer sets the initial conditions for a trial solution run.

The analog computer solves the system state and costate equations. In so doing, it calculates the extremal control and the cost, halting the computer run when the cost exceeds the specified terminal cost.

The digital computer is used for storage of the terminal results of a trial solution. The results are tested and further values calculated according to the iteration scheme in the digital computer.

Output of the final solution is done on the typewriter of the teletype unit.

3.3 Description of the Digital Program

Language and Speed

The program is written in an assembler language for the PDP-8. The DEC language PAL-D forms the basis for the program, with a modification of their floating point language being used for the arithmetic.^{[6],[7]} The low-level language is required by the necessity of rapid calculation and core manipulation. Some of the arguments for the use of low-level languages are presented in Bell and Griffin.^[4] The explanation of the program will be concerned with blocks of instructions, not with individual instructions, in order to reach the important points and to maintain continuity. For the rest of this section, reference to Appendix 4 will be required, especially to the flow chart, Figure A-4.1.

The Main Program

START

The input message is printed and the initialization executed to prepare for data input for a new case. The input data set is typed by the operator as a string of numbers in floating point representation (decimal) and each number is stored. This is done by the READ loop using indirect addressing and address modification. The initial states are set on the appropriate digital to analog converter channels.

ENTRY

An analog computer run is made by the subroutine RUN. The value NX of $||\underline{D}||$ at the terminal time is calculated by subroutine SNX.

NX thus is the distance from the terminal state to the target set. If NX is less than EX, the specified state error tolerance, execution continues at OUTPT, which will be described following this section. If not, the terminal error norm, ND, is calculated by subroutine SND. If ND is less than ER, the terminal error tolerance, execution continues at INCOST. If not, control is transferred to the iteration scheme composed of instructions at SET1, SET2, SET3 and SET4.

OUTPT

To expedite solution, a two-stage search has been programmed. The first time that OUTPT is reached, (INDEX=0), the tolerances and step sizes are decreased and the problem restarted from the current value of initial costate. The second time OUTPT is entered, when the changed terminal error tolerance has been reached, the final values are printed out and control returned to START for entry of a new case from the keyboard.

IN COST

IN COST is entered when the terminal matching condition is satisfied. The cost is increased by DD and control is returned to ENTRY for a new search for a costate vector.

The Iteration Scheme -- SET1, SET2, SET3, and SET4.

For this section to be reached, OUTPT and INCOST have not been entered.

SET1 is entered in the first step of an iteration, when ENTRY was reached from START or INCOST, and NU is -1. The costate is called the reference costate, (N1, N2) and the value of terminal error

norm ND is saved as DM, corresponding to (N1, N2). NU is set at 0, N1 set to N1 + DE and control returned to ENTRY.

Next, SET2 is entered. ND - DM is saved as DN1, corresponding to (N1 + DE, N2). NU is set at 1, N1 reset to the reference value, N2 set to N2 + DE, and control returned to ENTRY.

When SET3 is entered, N2 is reset to the reference value, ND - DM is saved as DN2, corresponding to (N1, N2 + DE), and NU is set at 2. The magnitude of the gradient of the terminal error norm is calculated as $M = (DN1^2 + DN2^2)^{1/2}/DE$ and the steps in the negative gradient direction are calculated as $(-DN1/M, -DN2/M)$. A new reference costate is calculated as $(N1 - DN1/M, N2 - DN2/M)$. Control is returned to ENTRY.

The final step of the iteration scheme is a repetition step, at SET4. If the value of DN is less than that previous (DM), DN is saved as DM and the correction procedure is repeated. The negative gradient steps are added to the reference costate forming a new reference costate and control is returned to ENTRY. When the value of DN returned is not less than DM, this repetition ends. The negative gradient steps are subtracted from the reference costate to give the previous (and better) value, NU is set at -1 to force the calculation of a new gradient, and control returned to ENTRY.

Subroutines

DACON

This subroutine converts the octal number in the accumulator into a voltage on the channel given by DACHN. The values -3777_8 to 3777_8 (-2047_{10} to 2047_{10}) are converted to voltages in the range -10 volts to 10 volts.

ADCON

This subroutine converts a voltage on the channel in the accumulator into an octal number in the accumulator, using the same scale as indicated in DACON.

RUN

This subroutine operates the system once. The cost and initial costates are digital-to-analog converted and set onto the initial condition channels. The analog computer is put into operate mode, after a pause for charging of the initial condition capacitors. Analog-to-digital channel 0 is then tested until the run-end condition has been met and the computer placed into the hold mode. The values of the terminal state and costate are then converted to digital values and stored.

SNX

This subroutine calculates the Euclidean norm of the vector \underline{D}_T as NX. In the case programmed here, the target set is the line segment on the X_1 axis with right and left boundaries, DL and IL, respectively. \underline{D}_T is calculated as

$$\underline{D}_T = G - \underline{x}_T = \begin{cases} (DL - X_1, -X_2) & X_1 > DL \\ (0, -X_2) & IL \leq X_1 \leq DL \\ (IL - X_1, -X_2) & X_1 \leq IL \end{cases} \quad (3.5)$$

The original value of X_1 is saved as X_{1R} , and X_1 is used for the first component of \underline{D}_T in the remainder of the subroutine and in SND.

$$\text{The norm of } \underline{D}_T \text{ is calculated as } NX = (D_1^2 + D_2^2)^{\frac{1}{2}}. \quad (3.6)$$

SND

This subroutine calculates the Euclidean norm of the terminal

costate vector. Then the terminal error norm is calculated as the value ND . This was given in (3.3)

3.4 Extensions of the Program

Higher Order Problems

The program could be conveniently extended to higher order systems than second order ones. The input, operating and output subroutines would simply require additional steps to carry the extra values to and from the analog computer. The norm calculations would be modified only by the addition of additional squared terms. A branch of the iteration scheme would be added to correspond to SET1 and SET2, where the steps are taken in calculating the gradient, in order to calculate each extra component of the gradient. The major foreseeable difficulty is in the analog program, where it has been found that special and possibly digital computer controlled approaches are necessary as the order of the system increases to handle scaling problems with the costate variables.^{[3],[9]}

Other Target Sets

Extension to other target sets which are fixed is a simple matter of modifying the subroutine SNX to properly calculate the value of \underline{D}_T . This is possible for the target sets discussed in section 2.2, which have a finite number of attracting regions, with difficulty directly dependent on the complexity of the target set.

The time-varying case will have only the added complexity of predicting the final time in order to use the methods suggested in section 2.3. SNX would be modified to predict the final time, or to use the current time to calculate the $||\underline{D}||$ value.

CHAPTER 4

APPLICATIONS

In this chapter, as the title indicates, the technique developed in the earlier chapters is applied to a linear second order system and a nonlinear second order system.

4.1 The Harmonic Oscillator

The undamped harmonic oscillator, extensively studied in the literature, was chosen as an example of a linear system. This system is described by the differential equation

$$\frac{d^2 x(t)}{dt^2} + \omega^2 x(t) = K u(t), \quad K \geq 0, \quad |u(t)| \leq 1. \quad (4.1)$$

Using the set of normalized state variables

$$\begin{aligned} \dot{x}_1(t) &= \frac{\omega^2}{K} x(t) \\ \dot{x}_2(t) &= \frac{\omega^2}{K} \dot{x}(t) \end{aligned} \quad (4.2)$$

the state equation becomes

$$\underline{\dot{x}}(t) = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \underline{x}(t) + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(t). \quad (4.3)$$

The problem of transition of the state from the initial state $\underline{x}(t_0)$ to a line segment on the x_1 axis in minimum time will be considered.

$$\underline{x}(t_0) = \underline{x}^{(0)} = \begin{bmatrix} x_1^{(0)} \\ x_2^{(0)} \end{bmatrix} \quad (4.4)$$

$$g = \{ \underline{g} \mid \underline{g} = \alpha \underline{g}_1 + (1 - \alpha) \underline{g}_2, \alpha \in [0, 1] \} \quad (4.5)$$

$$G(\underline{x}) = \{ \underline{x} \mid \| \underline{x} - \underline{g} \| \leq R \} \quad (4.6)$$

$$\underline{x}(t_f) = \underline{x}^{(f)} \in G(\underline{x}). \quad (4.7)$$

A detailed derivation of the exact solution to this problem is found in Athans and Falb^[1]. The Hamiltonian

$$H(\underline{x}, \underline{u}, \underline{p}) = 1 + p_1 x_2 - p_2 x_1 + p_2 u \quad (4.8)$$

is minimized by

$$u(t) = - \operatorname{sgn} p_2(t). \quad (4.9)$$

The costate equation is given by

$$\dot{\underline{p}}(t) = - \frac{\partial H}{\partial \underline{x}} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \underline{p}(t) \quad (4.10)$$

The following points can easily be shown to be true. The solution of the costate equation is simply that of an undamped, uncontrolled harmonic oscillator; the costate trajectories being circles about the origin. The time optimal control must be piecewise constant by (4.9). There can be an unlimited number of switchings, due to the harmonic nature of $p_2(t)$. There is no possibility of singular control.

In the solution using the origin as the target set, it can be shown that the extremal trajectories are circles about the point $(\Delta, 0)$ in state space, with $\Delta = \pm 1$ being the control specified by (4.9). The time taken to move the state along one of these circular arcs is given by the angle at the centre of the circle subtended by the arc. An analytical method of finding the time optimal solution is thus available. The angle sum gives the time, while the time to the first switching gives the initial costate.

The state and costate equations were programmed on the analog computer, with initial conditions supplied by the digital computer program. Unity scaling was used (within the arbitrary factor of $\frac{\omega^2}{K}$) so

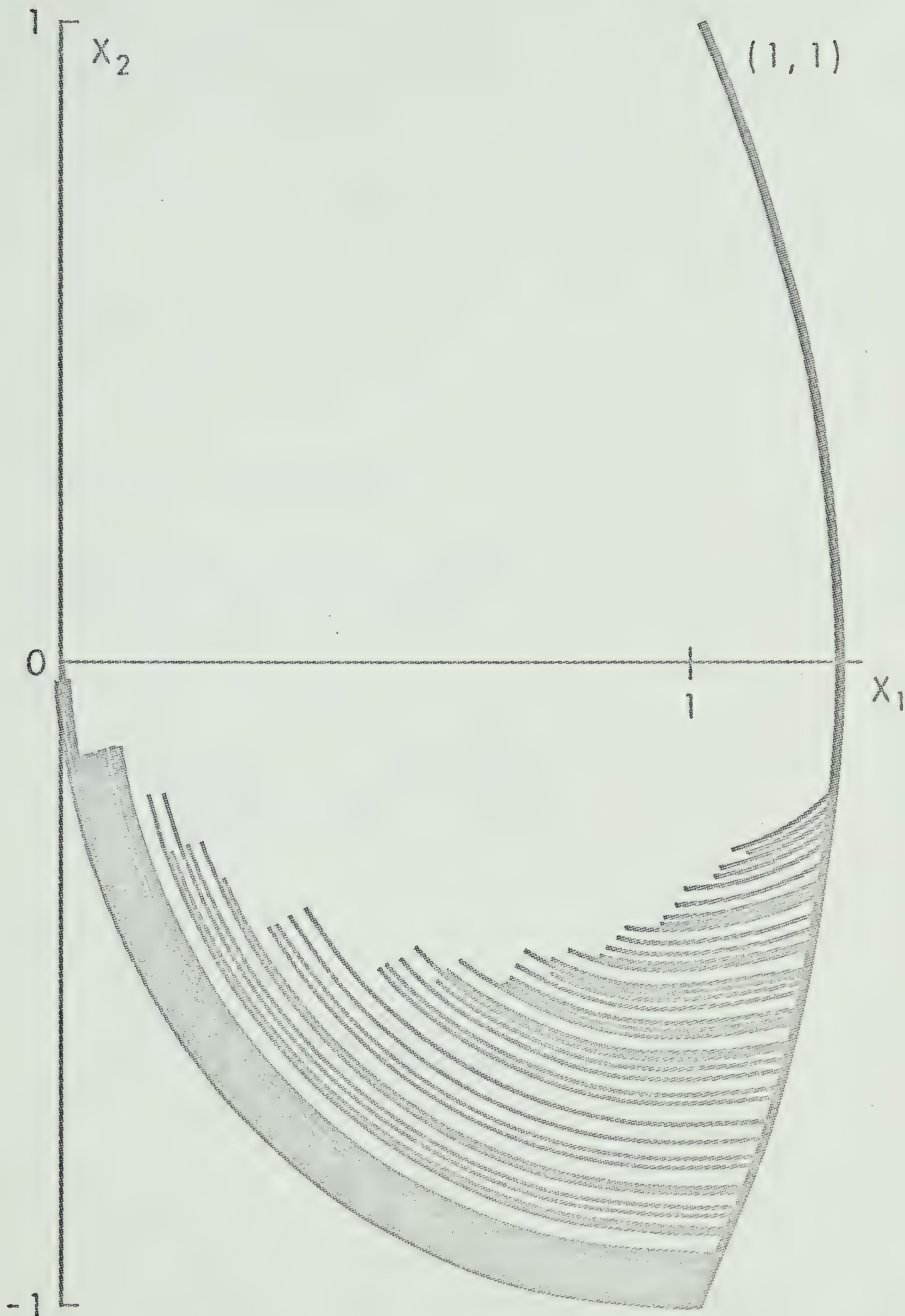


FIGURE 4.1 TYPICAL HARMONIC OSCILLATOR SOLUTION

HYERID COMPUTER SOLUTION

INITIAL: COST, STATES, COSTATES

PARAMETERS: RIGHT, LEFT TARGET POINTS, COST & COSTATE STEPS

ERROR NORMS: STATE & TERMINAL

0.0 1.0 1.0 -3.0 -3.0 0.0 0.0 0.1 0.1 0.1 0.2

FINAL: STATES + .0000 AND - .0097
 COSTATES - 2.4785 AND - 3.3939.
 COST + 2.4999 (SECONDS).
 + 122 ITERATIONS.

FIGURE 4.2 INPUT AND OUTPUT FOR FIGURE 4.1

that the problem was limited by the ± 10 volt range of the analog computer. One volt was used as one unit for the solution.

The solution of the optimal control problem was found for a variety of cases. The results of some of these tests are given in Table 4.1. For the cases with the origin as the target set, the exact solution is included. The time taken by the hybrid computer for solution, exclusive of teletype output, is included for some of the tests. Figure 4.1 shows a typical solution of the system. It is accompanied by its input and output, in Figure 4.2. The general characteristics of these results are discussed later in this chapter.

4.2 A Bilinear System

A bilinear system considered by Moon and Mohler^[11] was used as the second example. This nonlinear system is bilinear in the sense that it is linear in states and controls when separately considered. However, this property is not of any significance in the present discussion.

The problem is that of moving a searchlight and stopping at a specified angle in minimum time. The searchlight dynamics are controlled by the driving motor, which is controlled by armature current and by a brake. The losses in the system are negligible. The torque of the DC motor is proportional to the armature current.

The state equation is (from Moon and Mohler^[11])

$$\underline{x}(t) = \begin{bmatrix} x_2(t) \\ -a_2(1 + u_2(t))x_2(t) + a_1u_1(t) \end{bmatrix} \quad (4.11)$$

where $x_1(t)$ is the angular position in radians, $x_2(t)$ is the angular velocity in radians per second, $u_1(t)$ is the armature current, $u_2(t)$ is the braking control and a_1 and a_2 are appropriate positive constants.

TABLE 4.1

Minimum Time Solutions for the Second Order Harmonic Oscillator

Part A: The Target Set is the Origin

Experimental Results

Initial State	Initial Costate Guess	Solution Time (sec.)	Number of Iterations	Final Time (sec.)	Final Costate Ratio	Final Time (sec.)	Final Costate Ratio	Costate Error %
(1,1)	(-2,-2)	7.4	117	2.48	.7568	2.50	.7499	.92
		9.0	143		.7539			.53
	(-3,-3)	5.9	89		.7514			.20
		5.7	92		.7512			.17
	(-4,-4)	6.3	101		.7500			.01
		6.1	99		.7486			-.17
		6.1	100		.7490			-.12
		6.2	101		.7498			-.01
	(-5,-5)	6.5	108		.7491			-.11
		6.9	113		.7499			0
	(-6,-6)	8.0	130		.7505			.08
		7.8	127		.7481			-.24
(1,1)	(Average)	6.83	110	2.48	.7507	2.50	.7499	.10
(4,4)	(-4,-4)	39.8	544	8.98	.9330	8.99	.9289	.44
		20.0	309		.9325			.39
		21.0	307		.9342			.57
		15.6	231		.9363			.80
	(-5,-5)	17.2	250		.9356			.72
		19.4	284		.9347			.62
		18.2	267		.9363			.80
		12.8	192		.9362			.79
	(-6,-6)	16.6	240		.9319			.32
		18.0	257		.9335			.50
		16.0	234		.9365			.82
		15.8	233		.9352			.68
(4,4)	(Average)	19.4	279	8.98	.9347	8.99	.9289	.62
(4,4)	(Average 9 other runs)		381	8.98	.9341	8.99	.9289	.56
(5.405,0)	(-5,0)		222	8.30	7.30	8.32	6.73	8.48
			374	8.30	7.01			4.16
			134	8.32	7.16			6.40
(0,6.325)	(0,-6)		1766	9.74	.01278	9.73	.00931	37.27

TABLE 4.1

Part B: The Target Set is a Line Segment on the x_1 -axis

Experimental Results

Initial State	Initial Costate Guess	Target Segment	Final x_1 Value	Number of Iterations	Final Time (sec.)	Final Costate Ratio
(4,4)	(-3,-4)	[-1,1]	-.42	403	8.90	.807
		[-1,1]	-.703	553	8.90	.735
		[0,1]	0	640	8.98	.942
		[-.5,1]	-.5	365	8.90	.789
		[-.4,1]	-.4	953	8.92	.820
		[-2,1]	-.776	1179	8.90	.717

$$\text{For } i = 1, 2 \quad |u_i| \leq 1. \quad (4.12)$$

The initial state is given as $\underline{x}^{(o)}$ and the final state is required to be $(0, 0)^T$ in the original problem. The Hamiltonian is

$$H(\underline{x}, \underline{u}, \underline{p}) = 1 + p_1 x_2 + p_2 (a_2 (1 + u_2) x_2 + a_1 u_1) \quad (4.13)$$

and the costate is given by the equation

$$\underline{p}(t) = \begin{bmatrix} 0 \\ -p_1(t) + a_2(1 + u_2)p_2(t) \end{bmatrix}. \quad (4.14)$$

Minimization of the Hamiltonian with respect to the control yields the control functions

$$u_1(t) = -\text{sgn } a_1 p_2(t) = -\text{sgn } p_2(t) \quad (4.15)$$

$$u_2(t) = \text{sgn } a_2 p_2(t) x_2(t) = \text{sgn } p_2(t) x_2(t)$$

The problem was solved using the following values:

$$t_o = 0; a_1 = 2; a_2 = 1; x_1^{(o)} = 5; x_2^{(o)} = 10. \quad (4.16)$$

The state and costate equations were programmed on the analog computer, along with the control functions. The initial conditions were supplied by the digital computer program. Unity scaling was used, thus $x_2^{(o)}$ was set as close to full scale as possible, 9.997 volts, one bit short of 10 volts.

The case described was studied in depth, since the problem was almost full scale, and since comparison with the results of Moon and Mohler was possible. Cases using target sets other than the origin did not add any additional information to the study.

The results of some of the tests are recorded in Table 4.2. A typical solution is shown in Figure 4.3, with its input and output in Figure 4.4. The general characteristics of these results are discussed in the following section.

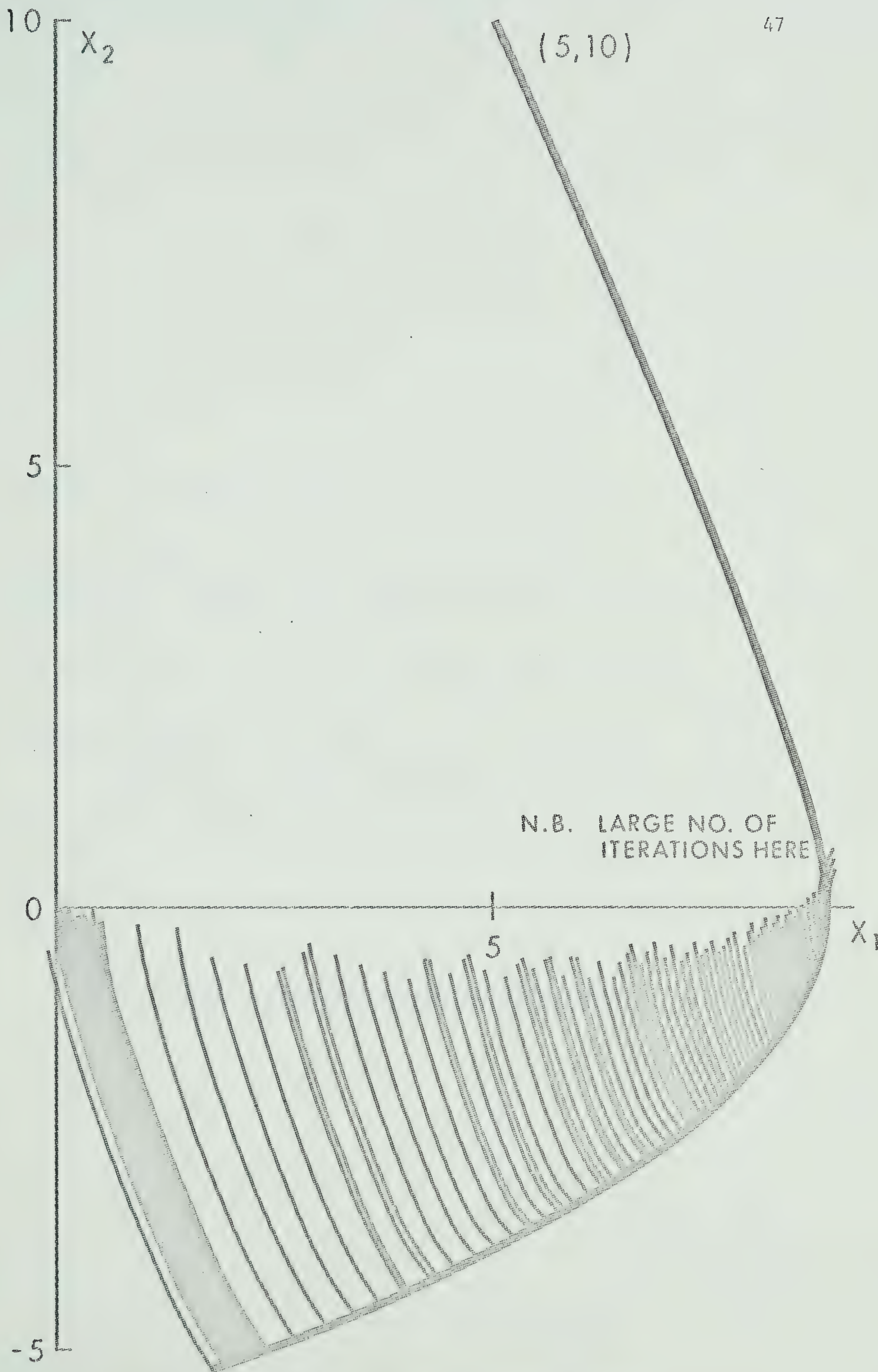


FIGURE 4.3 TYPICAL BILINEAR SYSTEM SOLUTION

HYBRID COMPUTER SOLUTION

INITIAL: COST, STATES, COSTATES

PARAMETERS: RIGHT, LEFT TARGET POINTS, COST & COSTATE STEPS

ERROR NORMS: STATE & TERMINAL

0.0 5.0 9.997 -3.0 -6.0 0.0 0.0 0.1 .025 .2 .15

FINAL: STATES + .0146 AND + .0097
 COSTATES - 1.3401 AND - 6.3496.
 COST + 4.9399 (SECONDS).
 + 615 ITERATIONS.

FIGURE 4.4 INPUT AND OUTPUT FOR FIGURE 4.3

TABLE 4.2

Minimum Time solutions for a Second Order Bilinear System

Experimental Results

Initial State	Initial Costate Guess	Target Segment	Number of Iterations	Final Time (sec.)	Final Costate Ratio
(5,10)	(-3,-6)	[0,0]	5077	4.90	.2132
			3235		.2123
			3518		.2132
			612		.2151
			713		.2132
			1146		.2128
			680		.2141
			514		.2141
			637		.2143
			1640		.2143
			4144		.2165
			1193		.2149
			2040		.2165
			1099		.2170
			1870		.2208
			434		.2228
			892		.2199
			2875		.2199
			810		.2199
			2889		.2248
			788		.2284
			738		.2304
			868		.2359
			984		.2357
			1216		.2361
			1061		.2558
			1104		.2456

4.3 Discussion of Results

The main objective of this research has been more to examine the effectiveness and validity of the technique developed rather than to compile solutions for a large number of examples. Consequently, only a few of the large number of cases treated were studied in depth. The general conclusions, however, are based upon the results of these detailed studies and upon the information gained by trying the other cases.

It is well known that the costate equation must be linear in the costate, so that the costate can be found only to within a scale factor with its sign determined. For this reason, the ratio of costates is reported here. In order to consider values in the full dynamic range of the costate to ensure that the more influential component was used half the time, both the costates were varied. This also tested iteration techniques for handling two simultaneous searches for future applications.

The tests on the harmonic oscillator showed that a rapid solution was indeed possible using the program developed to implement the technique. The solution times were in the order of ten seconds. The study of this system indicated the severity of step size effects. The use of large steps was found to result in fewer iterations until the target set was approached closely. Then the method would usually fail to converge, since the terminal states had become highly sensitive to initial costate changes and large steps were being used. On the other hand, small step sizes usually guaranteed convergence, but usually with a large number of iterations. These effects suggested the implementation of two stages of search, the first one with large steps until

the target set was approached closely, and the second one with smaller steps until the solution had converged sufficiently. This strategy was adopted, using the arbitrary factors of five for the cost and costate steps, and terminal condition error tolerance, and ten for the terminal error. In general, it was evident that step size choice was a compromise between speed of solution and risk of failure to converge.

The cases tested showed a lack of close repeatability. There were three factors in this. The solution accuracy in "fast-time" operation is probably closer to 1% than to .1%. This meant that the terminal values fluctuated from run to run. The discrete steps of the analog data after conversion to digital data was in effect quantization into steps which grew in proportion as the value became smaller. Consider a quantity which approaches zero. The number of bits representing it decreases, until each bit is a large part of the quantity. An analog voltage less than .01 volts in magnitude can be only one of five values: -2, -1, 0, 1, 2 (octal). The third factor was the affect of rounding the computed costate changes for digital to analog conversion. This caused steps in the initial costates rather than continuous changes.

One factor which appeared to enter into the cases studied in depth was the switching error. The $(1, 1)^T$ case had only one switching, whereas three switchings occurred in the $(4, 4)^T$ case. The finite speed of the comparator used for switching ($7\mu s$) probably contributed to the greater mean error in the costate ratio for the latter case.

The bilinear system showed a major difficulty in application to some problems. The progress of the search was arrested for a large number of iterations while an abrupt change in the initial costate was

made. This brought into question the assumption of Miura et al^[10], namely that the initial costate changes continuously with increasing cost. It became obvious that this assumption failed whenever the mode of solution being followed eventually led away from the target set, at which time a new mode for the control strategy had to be found. The program was able to find the new mode but the technique was not designed for jumps in the initial costate and thus did not handle them efficiently.

A wide range of solution times (not tabulated) was observed due to the jump in the initial costate, from 25 to 100 seconds. In the cases examined, a large number of iterations occurred, occupying a substantial portion of the solution time, while the initial costate was changed to allow continuation. This is noted in Figure 4.3.

Although difficult to compare directly, Moon and Mohler's results were available for comparison. They reported that the system was solved using their method with an IBM 360/75J at the University of California at Los Angeles in approximately 4.05 seconds per run, using less than 64K bytes (with 32 bits per byte).^[11] The PDP-8 used just over 1K bytes, with a little help from the TR-48 analog computer. It is probably safe to consider the cost of using the 360/75J as somewhat more than ten times that of the PDP-8-TR-48 hybrid, while the solution time was no more than ten times better.

CHAPTER 5

COMMENTS AND CONCLUSIONS

5.1 Improvements in the Program

The conclusions reached in this thesis are based somewhat on the experimental work, and are thus influenced by its imperfection. The possible improvements in the program are discussed here in order to moderate some of the adverse affects of the particular implementation of the technique.

The most important problem which depended on the program was concerned with the search strategy. With the "a posteriori" knowledge that jumps in the initial costate are possible, the program could employ a strategy which used very large steps at the appropriate time to find a new mode of extremal control. During normal searching on the terminal error condition, the steps used could be variable depending on the current success or failure of the gradient search, according to the planned strategy. It would be desirable to increase the step size whenever the convergence appears satisfactory and to decrease the step size if the steepest descent iterations are not succeeding and at the final stages of meeting the terminal error and target conditions.

The total execution time could be reduced by about twenty-five per cent by the use of machine language programming exclusively in the solution iterations. Since all values communicated can be only one word long, no significant use of the Floating Point System was made in the calculations. One word arithmetic, possibly with provision for

some round-off controls could just as well have been used. The programming effort would be increased, however.

5.2 Summary

The technique initiated by Miura, Tsuda and Iwata^[10] for the solution of optimal control problems on the hybrid computer has been examined and extended in this thesis. The algorithm for more general target sets was derived and shown to work on a second order system.

It was shown that the assumption, made in Miura et al^[10], that the extremal solution which is initially closest to the target set is the optimal one, does not always hold. A counter-example was described in Chapter 2 to support this statement. In addition, it was shown experimentally that the assumption that changes in initial costate are continuous with increasing cost is invalid, and it was to be expected that shifts or jumps in the initial costate would occur as cost increased. This was caused by the need for new modes of control which were not initially minimal-distance ones.

The hybrid computer program was shown to be usable on some examples, even when the unexpected shift in the initial costate appeared. The efficiency of the program could be improved for regular use of the program. The use of "fast-time" integration was necessary to make solution times reasonable. The comparison between the solution from Moon and Mohler^[11] and that in this thesis is difficult to make, but the product of capital cost and solution time for the hybrid computer is probably smaller than for the IBM 360/75J. Satisfactory solution of these systems on a PDP-8 alone would be slow or impossible because of the small memory available.

5.3 Suggestions for Future Work

The technique developed here could be used in the solution of optimal control of deterministic systems modelled on the analog computer. The validity of the algorithm for a particular system would have to be established because the algorithm is not always applicable. The convergence to an extremal solution could be guaranteed by improvements in the program to handle shifts in the initial costate.

The limitations of this technique seem to indicate that more work should be done in order to establish sufficient conditions for the solution of given classes of problems. Using the method on more complex systems without establishing its validity might prove futile. The application to time-varying target sets should be made. In the case of systems for which the assumptions made in this thesis are valid, the application of some other techniques, which do not use the costate and the minimum principle, such as perturbation techniques, should be examined.

Since it has been found that the technique works for the second order harmonic oscillator, the relationship between the ideal (undamped) case and non-ideal cases (e.g., positive and negative damping) could be examined.

The on-line control of a simple real system might be attempted by simulating the system on the hybrid computer, using the model to determine the optimal control strategy and then using the hybrid computer to apply the strategy to the real system.

It is proposed to submit the significant results in this thesis for publication in the IEEE Transactions on Computers.

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APPENDIX 1

Some Definitions and Results in Optimal Control Theory^[1]

1. A Dynamical System

A dynamical system S is the composite concept consisting of sets T , Σ , Ω , and U , a variable $\underline{x}(t)$, and a function \underline{g} such that the axioms listed are satisfied. T , a subset of the real numbers, is called the domain of the system. Σ is the (metric) state space of the system. U , the input space of the system is a set of piecewise continuous functions on T with values in the (metric) space Ω . $\underline{x}(t)$ is the state variable and is defined on T with values in Σ .

$(t_0, t]$ is a half-open interval in T and used as an argument denotes a segment of a function on T restricted to $(t_0, t]$. \underline{g} is a function from $\Sigma \times \Omega \times T$ into E^P . $\underline{u}(t_0, t]$ is a segment of $\underline{u}(t)$ in U , called the input over the observation interval to the system.

$\underline{y}(t_0, t] = \underline{g}[\underline{x}(t_0), \underline{u}(t_0, t)]$ is a corresponding output of the system, with $(\underline{u}(t_0, t], \underline{y}(t_0, t])$ forming an input-output pair of the system. $\underline{x}(t) = \underline{\phi}[t; \underline{u}(t_0, t], \underline{x}(t_0)]$ describes the trajectory of the system on $(t_0, t]$, starting from $\underline{x}(t_0)$ and generated by input $\underline{u}(t_0, t]$.

It is assumed that the state at t_0 , $\underline{x}(t_0)$, and the control $\underline{u}(t_0, t]$ specify a uniquely defined output. The system is nonanticipative since future values of the input do not affect $\underline{y}(t_0, t]$.

It is assumed that every input-output pair has a corresponding state history $\underline{x}(t_0, t]$ in Σ .

It is assumed that \underline{g} and $\underline{\phi}$ are smooth in their arguments so that small changes in the input cause small changes in the output and

trajectory of the system.

It is assumed that the transition function ϕ satisfies the transition, or semigroup, condition in that the initial condition corresponds to the starting point of the trajectory; and if the input takes the system from \underline{x}_0 to \underline{x} along a trajectory, and if $\hat{\underline{x}}$ is a state on the trajectory, then the input will take the state from $\hat{\underline{x}}$ to \underline{x} .

2. Finite Dimensional

A dynamical system S is finite dimensional if the state space is a Euclidean space, $\Sigma = E^n$; and Ω is a Euclidean space, $\Omega = E^m$, with $m \leq n$. n is the dimension of the system.

3. Continuous-Time

A dynamical system S is continuous-time if T is an open interval of the real numbers.

4. Differential

A dynamical system S is a differential system if the state is given by a solution of a differential equation system $\dot{\underline{x}}(t) = \underline{f}(\underline{x}, \underline{u}, t)$ $\underline{x}(t_0)$ as initial condition and \underline{g} is a continuous function of its arguments.

APPENDIX 2

Some Aspects of Hybrid Computation

Hybrid computation is discussed from the viewpoint of optimal automatic control studies with the plant simulated in the analog computer. This restricts the scope of this appendix, which is not intended as a tutorial account for the uninitiated, to the area of concern of this thesis. The object is to summarize some of the aspects of hybrid computation examined by the author and to indicate some generalizations which have been made.

Bekey and Karplus define a spectrum of hybrid computing techniques and systems, ranging from pure analog computers with digital logic to pure digital computers with analog inputs. They state, "The most extensive and powerful of existing hybrid computer systems are comprised of general-purpose digital computers and general-purpose analog computers."^[3] Bell and Griffin concur, terming this type of system "the most flexible hybrid computing facility".^[4]

Specifically, such a hybrid computer consists of the analog and digital computers mentioned, with a communication link between them. The function of the communication link is to transfer analog, digital and logical signals between the computers, making the appropriate conversions. It is this type of hybrid computer which is implied in this appendix.

A-2.1 Inter-Computer Communication

In a hybrid computer system the communication link often becomes the "weak link" in the chain. The number of channels of analog-to-digital and digital-to-analog conversion (hereinafter termed ADC and

DAC, respectively) which are available determine the amount of information which can be transferred between computers. The speed of multiplexing determines the rate of transfer (with fast languages). The availability of extra logic signal channels can simplify many facets of computation without tying up an information channel. Mode control of the analog computer falls into this category.

The time taken by the communication link to perform the operations required may introduce a variable and significant time lag into some computations. This time lag is directly proportional to the slowness of the digital computer language used. For instance, the applications in which the digital computer is used for calculations of nonlinearities for the analog computer solution are seriously affected, especially in "fast-time" operation of the analog computer.

The effect of the communication link in speed of operation can be lessened by avoiding its use in "critically timed" operations. An analog comparator switching in 7 μ s is obviously better than an ADC, a test of the value, and a DAC consuming about 70 μ s, when using the switching within the analog computation. In the same category of operation is the use of an incremental conversion ADC fed directly into the accumulator of the digital computer and tested as often as possible. This technique would require about 7 μ s in the PDP-8-TR-48 system, but possibly less in a digital computer with half the cycle time of 1.5 μ s of the PDP-8. In any event, it is clear that using the digital computer in ongoing analog calculation requires study of the effects of computation times, both fixed and variable. There are several applications, however, in which the digital computer results

are used only when the analog computer is in hold mode. This is the technique used in this thesis.

Communication is also affected by the discretization of continuous analog data by the ADC. This is further discussed in the following section of this appendix.

A-2.2 Accuracy

Accuracy requirements in hybrid computation are determined by a somewhat different set of conditions than those affecting digital computation. In particular, the solution of a set of differential equations programmed on an analog computer will converge to a real solution (of the system programmed); whereas one has no guarantee that a given method and step size will converge to a solution in a digital computer solution of the same set. The precision used in digital computation is made necessary to some extent by the methods chosen for rapid solution, at the price of guarantees of convergence.

In simulations of many physical systems, accuracy in the order of one per cent is completely acceptable. The non-exact nature of the differential equations, parameters, input signals and outputs precludes the use of six to sixteen significant figures which are often necessary in the solution of the system on a digital computer. The limit of accuracy for analog computers is .01%, but the high speed transistor amplifier models are limited to .1%. In high speed operation this may become 1% as switching times and capacitive effects become more significant.

Since the costate equations are unstable for stable state equations systems, the magnitude of the initial condition of the costate must in general be sufficiently small to prevent the costate

system from reaching the limits of the analog computer amplifiers; yet it must be as large as possible to cope with discretization problems.

The discretization of the communicated data has an important effect here. A 12 bit word implies a step size of .049% of the analog computer full scale (10 volts for most transistor systems). For a costate value which increases with a dynamic range of 100 over the solution time, this would make each step 5% of full scale, a significant discretization error.

The analog computer, DAC and ADC must be calibrated properly to ensure maximum accuracy. The nature of the applications considered in this thesis are such that drift of the analog component characteristics does not have a significant effect, since run times are very short, and initial conditions are reset from the DAC for each run. Along with calibration, loading effects must be considered since fast, large bandwidth amplifiers are used. The comparator output may be susceptible to loading error, for example.

The accuracy of calculations in the digital computer must be considered, but in applications such as found in this thesis, the calculation requirements are few and are more or less independent of previous runs. This precludes the buildup of round-off and numerical instability errors.

A-2.3 Speed

A fundamental consideration is speed of solution. The speed of solution of a hybrid computer method must be of the same order as a good digital computer method to be competitive. This requires the use of the fastest possible techniques in both the analog and digital

portions of the hybrid system.

Most modern analog computers are constructed with two normal speeds of integration. The "slow-time" mode uses integrators of unity gain as the basic unit. The "fast-time" mode uses a gain of 500 or 1000. Normal usage of this mode is in repetitive operation with initial condition and operate modes cycled from 10 to 1000 (or more) times per second. If the mode control is accessible, the digital computer can control the analog computer, replacing the timing unit, or individually controlling sets of integrators. For studies such as found in this thesis, the high speed mode must be used whenever possible to decrease run time.

One important consideration when the digital computer is used to control the analog runs is that enough time must be allowed for the initial condition capacitors to be charged to the specified levels. To set initial conditions to within .05% of full scale requires 7.6 time constants of the charging network. The finite switching speed of the comparators and the mode controls must also be taken into account. In many computers, the hold mode is switched by relays, which may necessitate the use of track-and-hold amplifiers to stop critical values at the instant desired, faster than the computer can be switched into hold mode.

The communication link is likely to constrain the speed of solution in applications in which conversions form a large part of the calculation. Some systems are set up with relay multiplexing, which is almost useless at high speed operation. Solid state multiplexing and low settling time are important in this type of calculation.

The digital computer speed begins to become significant when high speed analog computation is used. One will often have the choice of different levels and types of languages. These are summarized here, specifically with reference to the PDP-8. An important consideration is execution time, since the program will be repeated many times for one solution.

The slowest languages are the interpretive ones, such as FOCAL from DEC. The operating system interprets each command and each variable when encountered in execution, performs the command using any subroutines necessary, and then reads and interprets the next instruction. In FOCAL, 10 ms is the minimum amount of time required to read a command and execute it. ADC and DAC take about 40 ms. Obviously, such a language is too slow for rapid solutions.

Compiler languages are much faster in execution. A language such as FORTRAN has a compiler which changes the instructions into machine language commands, stores variables, checks for errors and produces an "object program" which is the actual set of machine language commands. These languages are usually somewhat inefficient in that the compiler is not able to use the techniques a human programmer might use, and in that every program must have allowances made for situations which may be encountered in different types of programs. In the two types just mentioned, programming effort is lowest. Instructions can be efficiently written, with the task of sorting out the machine language sequences necessary to execute them left to the compiler.

A low level language which is a compromise between the higher level languages so far mentioned and the very low level languages to be

mentioned is illustrated by DEC's Floating Point System.^[6] This language contains instructions analogous to assembly language instructions, which are assembled in the same manner. They allow the programmer to program floating point operations on numbers consisting of three words, one for the power of 2 and two for the fraction between .5 and 1.0. Thus 5.0 is stored as $0003\ 2400\ 0000_8$, or $.625_{10} \times 2^3$. Some arithmetic operations and functions are available and input and output routines are used, as supplied or added by the author. The PAL-D Assembler^[7] is used to assemble the symbolic instructions into machine language code, with most of the organization of the program already done by the programmer. Further discussion is contained in the third appendix.

The lowest level of programming language is represented by DEC's PAL-D Assembler in its basic form. At this level each machine language instruction is represented by a symbolic code. Variables are addressed as names, and addressing and assignment of space is usually the concern of the programmer. Some automatic assignment of addresses is possible. The programmer, through the exertion of a substantial amount of programming effort, may be able to develop a very efficient program as far as execution time is concerned. The programming effort required, once basic skills at assembly language programming are acquired, may be justified by the efficiency, but a compiler language should be available for many applications and especially for prototype programs. The Floating Point System must be considered to be an assembly language in this respect, differing from the simple assembly languages only in the availability of a number of subroutines for useful operations.

The discussion of computer languages has been oriented to the PDP-8, which was used by the author. Other small computers with 4K or 8K of core also have the same type of possibilities and most of the preceding discussion will apply.

One further point is of importance. The output from the hybrid computer should be considered as a factor in the utility of applications. High speed solutions may be displayed on a storage oscilloscope, with photographs providing the "hard" copy which may be desired. The monitoring of the solution with the oscilloscope allows operator intervention in the case of unforeseen problems. Should the hybrid computer be used in an application where numerical values are required, then the (very) slow teletype or electric typewriter output should be used. In applications such as those in this thesis, the amount of time for printing out the few results is of the same order of magnitude as the solution time. Clearly, in a situation where the hybrid computer is used to solve a control strategy problem for a real-time system, the result should be immediately applied to the system, after which the pertinent permanent record should be made, possibly on a shared time basis with the real-time control program.

APPENDIX 3

Description of the Hybrid Computer

The hybrid computer used for this work is located in the Department of Electrical Engineering at the University of Alberta. It consists of one TR-48 solid-state analog computer, an interface, and one PDP-8 solid-state digital computer. The computer is operated in an "open-shop" manner.

The TR-48 is a medium capacity general-purpose analog computer, manufactured by Electronics Associated, Incorporated, West Long Branch, N.J. Its range is from -10 to + 10 volts with chopper-stabilized transistor operational amplifiers. The computing components consist of the following: 18 integrators, 36 summers, 5 electronic bipolar multipliers, 4 electronic comparators, 8 electronic switches, 60 pots, 8 servo-set pots, 8 feedback limiters and 2 twenty-segment diode function generators.

The mode of each pair of integrators can be controlled individually. Mode control is effected by electronic switching between operate mode and initial condition mode and by relay switching between these modes and hold mode. In repetitive operation mode, the mode control logic signals are cycled between operate and initial condition, with ten milliseconds initial condition mode and from ten to five hundred milliseconds for operate mode. This particular cycle can be replaced by any other scheme desired using the digital computer for control. It is this type of operation which is referred to as "fast-time", since in repetitive operation (whether or not internally cycled) the speed of the integrators is increased by five hundred. "Fast-time" operation

allows the high speed solution of differential equations, and makes the iterative solution of ordinary differential equations practical.

The interface or communication link is based on Redcor Series 610 Linkage System.

Twelve digital bits are converted to ± 10 volts of analog information by one of sixteen digital to analog converters. An analog voltage from one of twenty-four channels converted to thirteen bits (twelve being used) by the analog to digital converter by means of a solid-state multiplexer.

The remainder of the interface consists of logical and timing facilities designed and implemented in the Department of Electrical Engineering. The only one of concern here is the analog mode control facility, which translates the command from the digital computer into the appropriate logical values to bring the appropriate mode.

The PDP-8 is a small general-purpose digital computer manufactured by Digital Equipment Corporation, Maynard, Mass. It has 4K (4096) twelve bit words of core memory, operating with a cycle time of 1.5 μ s, two TU-55 Dectape magnetic tape units and a 32K disc memory. It has a full complement of hardware machine language instructions and can service sixty-four peripheral devices requiring three commands each. The analog mode control commands, and the commands used in ADC and DAC are examples of these. The PDP-8 has the extended arithmetic option, which improves arithmetic calculation speed by the use of hard wired instruction. An ASR-33 teletype provides teletype input and typewriter output to the computer by means of appropriate programs.

As important as the hybrid computer hardware is, the software must be flexible in order to make use of it. The PDP-8 is supplied with

a wide range of software from assembly to compiler to interpretive languages. In this thesis the PAL-D Assembly Language^[7] is used to assemble the machine language sections of the program. The reader is also referred to Introduction to Programming; Digital Equipment Corporation's general manual on programming the PDP-8 family of computers.

The Floating Point System^[6] has a number of features which made it convenient to use. The program, as supplied by DEC, allows operations to be performed on three-word floating point numbers. The first word is the power of two and the other two form a twenty-four bit mantissa between one half and one in magnitude when normalized. Input and output routines are provided for the teletype. Allowance is made for the addition of subroutines which operate on the floating accumulator (FAC). The operations used in this thesis are the following:

FADD A (FSUB A) - Add (subtract) the number A to the FAC, normalize and leave in the FAC.

FMPY A, (FDIV A) - Multiply (divide) the FAC by the number A, normalize and leave in the FAC.

FGET A - Load the number A into the FAC.

FPUT A - Store the FAC into the three words starting at A.

FNOR A - Normalize the mantissa of the FAC to magnitude between .5 and 1.0.

These are supplemented by the following subroutine-type commands:

FEXT - Leave floating interpreter. Instructions following are in machine language.

SQUARE - Square the FAC.

SQROOT - Take square root of the FAC.

FIX - Convert the FAC into a one-word number at location 45.

INPUT - Read a decimal number from the teletype.

OUTPUT - Print the contents of the FAC on the teletype as a decimal number.

CRLF - Print a carriage-return and line-feed.

NEWOUT - Execute OUTPUT, then CRLF.

RNDFIX - Convert the floating accumulator contents into the nearest one word number at location 45.

Three additional subroutines are used in the program. The subroutine called by FLOAT will transform the contents of the real accumulator into a normalized floating point number. The subroutine called by MSSGE will print, on the teletype, the text stored between the first address, immediately following, and the last address, whose negative follows the first address. The message is stored two literal characters to a word by the TEXT pseudo-operator of PAL-D. The character (\) is used to execute the instruction CRLF as the message is printed. The other subroutine used is called by IN, and is the entry into the floating point interpreter. All instructions following are interpreted as floating point ones, until the command FEXT is encountered.

The speed of execution of the floating point system lies between machine language programming and interpreter language programming. The interpretation of commands is what makes the difference. The arithmetic commands (FADD, FSUB, FMPY, FDIV) take from 330 to 385 μ s

to execute. In the interpreter language FOCAL, the equivalent commands take 20 to 40 ms. To set one variable equal to another (FGET A; FPUT B) takes 270 μ s. In FOCAL this would take about 20 ms. In this program, an ADC takes 69 μ s, and a DAC 44 μ s. Using the FOCAL functions takes about 40 ms. The use of the machine language and floating point programming increases the speed of digital operations by a factor of about 500. The net effect of high-speed operation is a speed increase of 500 times. Use of only one of the techniques (i.e. using normal speed integration or FOCAL programming) would have resulted in only two to five times the speed of normal operation.

APPENDIX 4 FEF-8 PROGRAM

SYMBOLS AND ABSOLUTE LOCATIONS

ADADD	1106	NU	0144
ADCON	0161	NX	1345
ADDR	0023	N1	0111
ADEND	1107	N2	0114
ADST	1060	ONE	1110
C	0100	OUTPT	0600
CHANGE	0253	PAUS	0032
CONST	0026	P1	1356
CONT	0505	P2	1361
DACHN	0160	READ	0214
DACON	0146	RUN	1000
DD	0125	SET1	0400
DE	0130	SET2	0411
DI	0145	SET3	0426
DL	0117	SET4	0465
DM	0553	SND	1252
DN1	0556	SNX	1200
DN2	0561	SNXC	1237
END1	0024	START	0201
ENTRY	0255	S1	1400
ER	0136	S2	1521
ERROR	0031	S3	1534
EX	0133	S4	1541
FINAL	0624	S5	1555
FIVE	0713	S6	1567
FLAG	1046	S7	1577
HUNDRED	0025	S8	1600
IL	0122	S9	1612
INCOST	0520	TEMP	0021
INDEX	0033	TEN	0716
IT	0034	THREE	0022
I1	0103	T1	1334
I2	0106	T2	1337
M	0564	UPPER	1230
MIDDLE	1223	X1	1350
M1	0020	X1R	1331
ND	0141	X2	1353
NP	1342		

EQUIVALENCE SPECIFICATIONS

FADD=1000
FSUB=2000
FMPLY=3000
FDIV=4000
FGET=5000
FPUT=6000
FNCR=7000
FEXT=0
SQUARE=1
SQROOT=2
FSIN=3
FCOS=4
ARCTAN=5
FLOG=6
FEXP=7
FIX=10
NEGFAC=11
NEGOP=12
INPUT=13
OUTPUT=14
CRLF=15
NEWOUT=16
RNDFIX=17
FLOAT=4405
MSSGE=4406
IN=4407
FIXTAB
C=100
I1=103
I2=106
N1=111
N2=114
DL=117
IL=122
DD=125
DE=130
EX=133
ER=136
ND=141
X1R=1331
T1=1334
T2=1337
NP=1342
NX=1345
X1=1350
X2=1353
P1=1356
P2=1361
DM=553
DN1=556
DN2=561
M=564

PROGRAM LISTING

		*5		
0005	4552		4552	
0006	4400		4400	/FLOAT SUBROUTINE ENTRY
0007	5600		5600	/MESSAGE PRINTOUT ENTRY
				/FLOATING POINT ENTRY
				/GENERAL QUANTITIES
		*20		
0020	7777	M1,	-1	
0021	0000	TEMP,	0	
0022	0003	THREE,	3	
0023	0100	ADDR,	100	
0024	7637	END1,	-141	
0025	0100	HUNDRD,	100	
0026	0010	CONST,	0010	
0027	3146		3146	
0030	3147		3147	
0031	7730	ERROR,	-50	
0032	4000	PAUS,	4000	
0033	0000	INDEX,	0	
0034	0000	IT,	0	
0035	0000		0	
0036	0000		0	
		*55		
0055	7777		7777	
0056	7777		7777	
		*144		
0144	0000	NU,	0	
0145	0000	DI,	0	
0146	0000	DACON,	0	/DIGITAL TO ANALOG
0147	6462		6462	
0150	6464		6464	
0151	7200		CLA	
0152	1160		TAD DACHN	
0153	6451		6451	
0154	6454		6454	
0155	6441		6441	
0156	5155		JMP .-1	
0157	5546		JMP I DACON	
0160	0000	DACHN,	0	/DAC CHANNEL
0161	0000	ADCON,	0	/ANALOG TO DIGITAL
0162	6412		6412	
0163	6421		6421	
0164	6433		6433	
0165	6411		6411	
0166	5165		JMP .-1	
0167	6421		6421	
0170	6422		6422	
0171	6411		6411	
0172	5171		JMP .-1	
0173	6421		6421	
0174	7200		CLA	
0175	6434		6434	
0176	5561		JMP I ADCON	

0200	7402		HLT	
0201	6046	START,	TLS	/START OF MAIN PROGRAM
0202	4406		MSSGE	/INPUT MESSAGE
0203	1400		S1	
0204	6257		-S2	
0205	7200		CLA	
0206	3033		DCA INDEX	
0207	3034		DCA IT	
0210	3035		DCA IT+1	
0211	3036		DCA IT+2	
0212	1025		TAD HUNDRD	/INPUT SEQUENCE
0213	3023		DCA ADDR	
0214	4407	READ,	IN	
0215	0013		INPUT	
0216	6423		FPUT 1 ADDR	
0217	0000		FEXT	
0220	3021		DCA TEMP	
0221	1022		TAD THREE	
0222	1023		TAD ADDR	
0223	3023		DCA ADDR	
0224	1023		TAD ADDR	
0225	1024		TAD END1	
0226	7440		SZA	
0227	5214		JMP READ	
0230	7201		CLA IAC	/SET INITIAL STATES
0231	3160		DCA DACHN	
0232	4407		IN	
0233	5103		FGET 11	
0234	3026		FMPY CONST	
0235	0017		RNDFIX	
0236	0000		FEXT	
0237	7200		CLA	
0240	1045		TAD 45	
0241	4146		JMS DACON	
0242	2160		ISZ DACHN	
0243	4407		IN	
0244	5106		FGET 12	
0245	3026		FMPY CONST	
0246	0017		RNDFIX	
0247	0000		FEXT	
0250	7200		CLA	
0251	1045		TAD 45	
0252	4146		JMS DACON	
0253	7240	CHANGE,	CLA CMA	/TOLERANCES CHANGED
0254	3144		DCA NU	
0255	4777	ENTRY,	JMS RUN	/OPERATE THE SYSTEM
0256	4776		JMS SNX	/TERMINAL ERROR TEST
0257	4407		IN	
0260	5775		FGET NX	
0261	2133		FSUB EX	
0262	0000		FEXT	
0263	1045		TAD 45	
0264	7710		SPA CLA	

0265	5774	JMP CUIPT	
0266	4773	JMS SND	/TERMINAL MATCHING TEST
0267	4407	IN	
0270	5141	FGET ND	
0271	2136	FSUB ER	
0272	0000	FEXT	
0273	1045	TAD 45	
0274	7710	SPA CLA	
0275	5772	JMP INCOST	
0276	1144	TAD NU	/ITERATION SCHEME DECISIONS
0277	7450	SNA	
0300	5771	JMP SET2	
0301	7510	SPA	
0302	5770	JMP SET1	
0303	1020	TAD M1	
0304	7550	SPA SNA	
0305	5767	JMP SET3	
0306	5766	JMP SET4	
0366	0465		
0367	0426		
0370	0400		
0371	0411		
0372	0520		
0373	1252		
0374	0600		
0375	1345		
0376	1200		
0377	1000		
	PAGE		
0400	4407	SET1, IN	/FIRST COSTATE STEP
0401	5111	FGET N1	
0402	1130	FADD DE	
0403	6111	FPUT N1	
0404	5141	FGET ND	
0405	6353	FPUT DM	
0406	0000	FEXT	
0407	3144	DCA NU	
0410	5777	JMP ENTRY	
0411	4407	SET2, IN	/SECOND COSTATE STEP
0412	5111	FGET N1	
0413	2130	FSUB DE	
0414	6111	FPUT N1	
0415	5114	FGET N2	
0416	1130	FADD DE	
0417	6114	FPUT N2	
0420	5141	FGET ND	
0421	2353	FSUB DM	
0422	6356	FPUT DN1	
0423	0000	FEXT	
0424	2144	ISZ NU	
0425	5777	JMP ENTRY	
0426	4407	SET3, IN	/STEEPEST DESCENT STEP
0427	5114	FGET N2	
0430	2130	FSUB DE	

0431	6114		FPUT N2	
0432	5141		FGET ND	
0433	2353		FSUB DM	
0434	6361		FPUT DN2	
0435	5356		FGET DN1	
0436	0001		SQUARE	
0437	6364		FPUT M	
0440	5361		FGET DN2	
0441	0001		SQUARE	
0442	1364		FADD M	
0443	0002		SQRCOT	
0444	4130		FDIV DE	
0445	6364		FPUT M	
0446	5356		FGET DN1	
0447	4364		FDIV M	
0450	6356		FPUT DN1	
0451	5361		FGET DN2	
0452	4364		FDIV M	
0453	6361		FPUT DN2	
0454	5111		FGET N1	
0455	2356		FSUB DN1	
0456	6111		FPUT N1	
0457	5114		FGET N2	
0460	2361		FSUB DN2	
0461	6114		FPUT N2	
0462	0000		FEXT	
0463	2144		ISZ NU	
0464	5777		JMP ENTRY	
0465	4407	SET4,	IN	/REPEATED STEEPEST DESCENT
0466	5141		FGET ND	
0467	2353		FSUB DM	
0470	0000		FEXT	
0471	1045		TAD 45	
0472	7710		SPA CLA	
0473	5305		JMP CONT	
0474	4407		IN	/REVERSE LAST STEP
0475	5111		FGET N1	
0476	1356		FADD DN1	
0477	6111		FPUT N1	
0500	5114		FGET N2	
0501	1361		FADD DN2	
0502	6114		FPUT N2	
0503	0000		FEXT	
0504	5776		JMP CHANGE	
0505	4407	CONT,	IN	/CONTINUE REPEATED STEPS
0506	5111		FGET N1	
0507	2356		FSUB DN1	
0510	6111		FPUT N1	
0511	5114		FGET N2	
0512	2361		FSUB DN2	
0513	6114		FPUT N2	
0514	5141		FGET ND	
0515	6353		FPUT DM	
0516	0000		FEXT	

0517	5777		JMP ENTRY	
0520	4407	IN COST,	IN	/TERMINAL CONDITION MET. /COST INCREASED
0521	5100		FGET C	
0522	1125		FADD DD	
0523	6100		FPUT C	
0524	0000		FEXT	
0525	5776		JMP CHANGE	
0576	0253			
0577	0255			
		PAGE		
0600	7200	OUTPT,	CLA	
0601	1033		TAD INDEX	
0602	7440		SZA	
0603	5224		JMP FINAL	
0604	2033		ISZ INDEX	
0605	4407		IN	/CHANGE TOLERANCES
0606	5125		FGET DD	
0607	4313		FDIV FIVE	
0610	6125		FPUT DD	
0611	5130		FGET DE	
0612	4313		FDIV FIVE	
0613	6130		FPUT DE	
0614	5133		FGET EX	
0615	4316		FDIV TEN	
0616	6133		FPUT EX	
0617	5136		FGET ER	
0620	4313		FDIV FIVE	
0621	6136		FPUT ER	
0622	0000		FEXT	
0623	5777		JMP CHANGE	
0624	7200	FINAL,	CLA	
0625	1376		TAD (10	
0626	3062		DCA 62	
0627	1375		TAD (4	
0630	3063		DCA 63	
0631	6046		TLS	/OUTPUT FINAL STATES
0632	4406		MSSGE	
0633	1521		S2	
0634	6244		-S3	
0635	4407		IN	
0636	5774		FGET X1R	
0637	0014		OUTPUT	
0640	0000		FEXT	
0641	4406		MSSGE	
0642	1534		S3	
0643	6237		-S4	
0644	4407		IN	
0645	5773		FGET X2	
0646	0014		OUTPUT	
0647	0000		FEXT	
0650	4406		MSSGE	
0651	1541		S4	
0652	6223		-S5	

0653	4407		IN	
0654	5111		FGET N1	
0655	0014		OUTPUT	
0656	0000		FEXT	
0657	4406		MSSGE	
0660	1534		S3	
0661	6237		-S4	
0662	4407		IN	
0663	5114		FGET N2	
0664	0014		OUTPUT	
0665	0000		FEXT	
0666	4406		MSSGE	
0667	1555		S5	
0670	6211		-S6	
0671	4407		IN	
0672	5100		FGET C	
0673	0014		OUTPUT	
0674	0000		FEXT	
0675	4406		MSSGE	
0676	1567		S6	
0677	6201		-S7	
0700	7200		CLA	
0701	3063		DCA 63	
0702	2062		ISZ 62	
0703	4407		IN	
0704	5034		FGET IT	
0705	0014		OUTPUT	
0706	0000		FEXT	
0707	4406		MSSGE	
0710	1600		S8	
0711	6166		-S9	
0712	5772		JMP START	
0713	0003	FIVE,	3	
0714	2400		2400	
0715	0000		0	
0716	0004	TEN,	4	
0717	2400		2400	
0720	0000		0	
0772	0201			
0773	1353			
0774	1331			
0775	0004			
0776	0010			
0777	0253			
		PAGE		
1000	0000	RUN,	0	/OPERATE THE SYSTEM
1001	7200		CLA	/SET COST
1002	3160		DCA DACHN	
1003	4407		IN	
1004	5100		FGET C	
1005	3026		FMPY CONST	
1006	0017		RNDFIX	
1007	0000		FEXT	

1010	7200		CLA	
1011	1045		TAD 45	
1012	4146		JMS DACON	
1013	2160		ISZ DACHN	/SET INITIAL COSTATES
1014	2160		ISZ DACHN	
1015	2160		ISZ DACHN	
1016	4407		IN	
1017	5111		FGET N1	
1020	3026		FMPY CONST	
1021	0017		RNDFIX	
1022	0000		FEXT	
1023	7200		CLA	
1024	1045		TAD 45	
1025	4146		JMS DACON	
1026	2160		ISZ DACHN	
1027	4407		IN	
1030	5114		FGET N2	
1031	3026		FMPY CONST	
1032	0017		RNDFIX	
1033	0000		FEXT	
1034	7200		CLA	
1035	1045		TAD 45	
1036	4146		JMS DACON	
1037	6322		6322	/INITIAL CONDITION MODE
1040	7200		CLA	/PAUSE, CHARGE CAPACITORS
1041	1032		TAD PAUS	
1042	3023		DCA ADDR	
1043	2023		ISZ ADDR	
1044	5243		JMP .-1	
1045	6321		6321	/OPERATE MODE
1046	7200	FLAG,	CLA	/TEST RUN-END CONDITION
1047	4161		JMS ADCON	
1050	1031		TAD ERROR	
1051	7510		SPA	
1052	5246		JMP FLAG	
1053	6324		6324	/HOLD MODE
1054	7200		CLA	/ADC TERMINAL VALUES
1055	3160		DCA DACHN	
1056	1306		TAD ADADD	
1057	3023		DCA ADDR	
1060	2160	ADST,	ISZ DACHN	
1061	1160		TAD DACHN	
1062	4161		JMS ADCON	
1063	4405		FLOAT	
1064	4407		IN	
1065	4026		FDIV CONST	
1066	6423		FPUT 1 ADDR	
1067	0000		FEXT	
1070	7200		CLA	
1071	1023		TAD ADDR	
1072	1022		TAD THREE	
1073	3023		DCA ADDR	
1074	1023		TAD ADDR	
1075	1307		TAD ADEND	

1076	7640		SZA CLA	
1077	5260		JMP ADST	
1100	4407		IN	
1101	5034		FGET IT	
1102	1310		FADD ONE	
1103	6034		FPUT IT	
1104	0000		FEXT	
1105	5600		JMP I RUN	
1106	1350	ADADD,	1350	
1107	6414	ADEND,	-1364	
1110	0001	ONE,	1	
1111	2000		2000	
1112	0000		0	
		PAGE		
1200	0000	SNX,	0	/FIND STATE ERROR NORM
1201	4407		IN	
1202	5350		FGET X1	
1203	6331		FPUT X1R	
1204	5350		FGET X1	
1205	2117		FSUB DL	
1206	6334		FPUT T1	
1207	0000		FEXT	
1210	1335		TAD T1+1	
1211	7500		SMA	
1212	5230		JMP UPPER	
1213	4407		IN	
1214	5350		FGET X1	
1215	2122		FSUB 1L	
1216	6334		FPUT T1	
1217	0000		FEXT	
1220	1335		TAD T1+1	
1221	7510		SPA	
1222	5230		JMP UPPER	
1223	7200	MIDDLE,	CLA	
1224	3350		DCA X1	
1225	3351		DCA X1+1	
1226	3352		DCA X1+2	
1227	5237		JMP SNXC	
1230	7200	UPPER,	CLA	
1231	1334		TAD T1	
1232	3350		DCA X1	
1233	1335		TAD T1+1	
1234	3351		DCA X1+1	
1235	1336		TAD T1+2	
1236	3352		DCA X1+2	
1237	4407	SNXC,	IN	
1240	5350		FGET X1	
1241	0001		SQUARE	
1242	6345		FPUT NX	
1243	5353		FGET X2	
1244	0001		SQUARE	
1245	1345		FADD NX	
1246	0002		SQROOT	
1247	6345		FPUT NX	

1250	0000		FEXT	
1251	5600		JMP I SNX	
1252	0000	SND,	0	/FIND TERMINAL ERROR NORM
1253	4407		IN	
1254	5356		FGET P1	
1255	0001		SQUARE	
1256	6342		FPUT NP	
1257	5361		FGET P2	
1260	0001		SQUARE	
1261	1342		FADD NP	
1262	0002		SQR00T	
1263	6342		FPUT NP	
1264	5350		FGET X1	
1265	4345		FDIV NX	
1266	6334		FPUT T1	
1267	5356		FGET P1	
1270	4342		FDIV NP	
1271	2334		FSUB T1	
1272	6334		FPUT T1	
1273	5353		FGET X2	
1274	4345		FDIV NX	
1275	6337		FPUT T2	
1276	5361		FGET P2	
1277	4342		FDIV NP	
1300	2337		FSUB T2	
1301	6337		FPUT T2	
1302	5337		FGET T2	
1303	0001		SQUARE	
1304	6141		FPUT ND	
1305	5334		FGET T1	
1306	0001		SQUARE	
1307	1141		FADD ND	
1310	0002		SQR00T	
1311	6141		FPUT ND	
1312	0000		FEXT	
1313	5652		JMP I SND	

		PAGE			
1400	0000	S1,	0	1420	1611 NI
1401	3434	TEXT Z\		1421	2411 TI
1402	1031	HY		1422	0114 AL
1403	0222	ER		1423	7240 :
1404	1104	ID		1424	4003 C
1405	4003	C		1425	1723 OS
1406	1715	OM		1426	2454 T,
1407	2025	PU		1427	4023 S
1410	2405	TE		1430	2401 TA
1411	2240	R		1431	2405 TE
1412	2317	SO		1432	2354 S,
1413	1425	LU		1433	4003 C
1414	2411	TI		1434	1723 OS
1415	1716	ON		1435	2401 TA
1416	3434	\		1436	2405 TE
1417	3411	\I		1437	2334 S\

1440	0000	Z		
1441	2001	TEXT	ZPA	
1442	2201	RA		
1443	1505	ME		
1444	2405	TE		
1445	2223	RS		
1446	7240	:		
1447	4022	R		
1450	1107	IG		
1451	1024	HT		
1452	5440	,		
1453	1405	LE		
1454	0624	FT		
1455	4024	T		
1456	0122	AR		
1457	0705	GE		
1460	2440	T		
1461	2017	PO		
1462	1116	IN		
1463	2423	TS		
1464	5440	,		
1465	0317	CO		
1466	2324	ST		
1467	4046	&		
1470	4003	C		
1471	1723	OS		
1472	2401	TA		
1473	2405	TE		
1474	4023	S		
1475	2405	TE		
1476	2023	PS		
1477	0000	Z		
1500	3405	TEXT	Z\N	
1501	2222	RR		
1502	1722	OR		
1503	4016	N		
1504	1722	OR		
1505	1523	MS		
1506	7240	:		
1507	4023	S		
1510	2401	TA		
1511	2405	TE		
1512	4046	&		
1513	4024	T		
1514	0522	ER		
1515	1511	MI		
1516	1601	NA		
1517	1434	L\		
1520	3400	\Z		
1521	0000	S2,		0
1522	3434	TEXT	Z\N	
1523	0611	FI		
1524	1601	NA		
1525	1472	L:		
1526	4040			
1527	2324	ST		
1530	0124	AT		
1531	0523	ES		
1532	4040			
1533	0000	Z		
1534	0000	S3,		0
1535	4040	TEXT	Z	
1536	0116	AN		
1537	0440	D		
1540	4000	Z		
1541	0000	S4,		0
1542	3440	TEXT	Z\N	
1543	4040			
1544	4040			
1545	4040			
1546	4040			
1547	0317	CO		
1550	2324	ST		
1551	0124	AT		
1552	0523	ES		
1553	4040			
1554	0000	Z		
1555	0000	S5,		0
1556	5634	TEXT	Z.\N	
1557	4040			
1560	4040			
1561	4040			
1562	4040			
1563	0317	CO		
1564	2324	ST		
1565	4040			
1566	0000	Z		
1567	0000	S6,		0
1570	4040	TEXT	Z	
1571	5023	CS		
1572	0503	EC		
1573	1716	ON		
1574	0423	DS		
1575	5156).		
1576	3400	\Z		
1577	0000	S7,		0
		PAGE		
1600	0000	S8,		0
1601	4040	TEXT	Z	
1602	1124	IT		
1603	0522	ER		
1604	0124	AT		
1605	1117	IO		
1606	1623	NS		
1607	5634	.\		
1610	3434	\N		
1611	0000	Z		
1612	0000	S9,		0

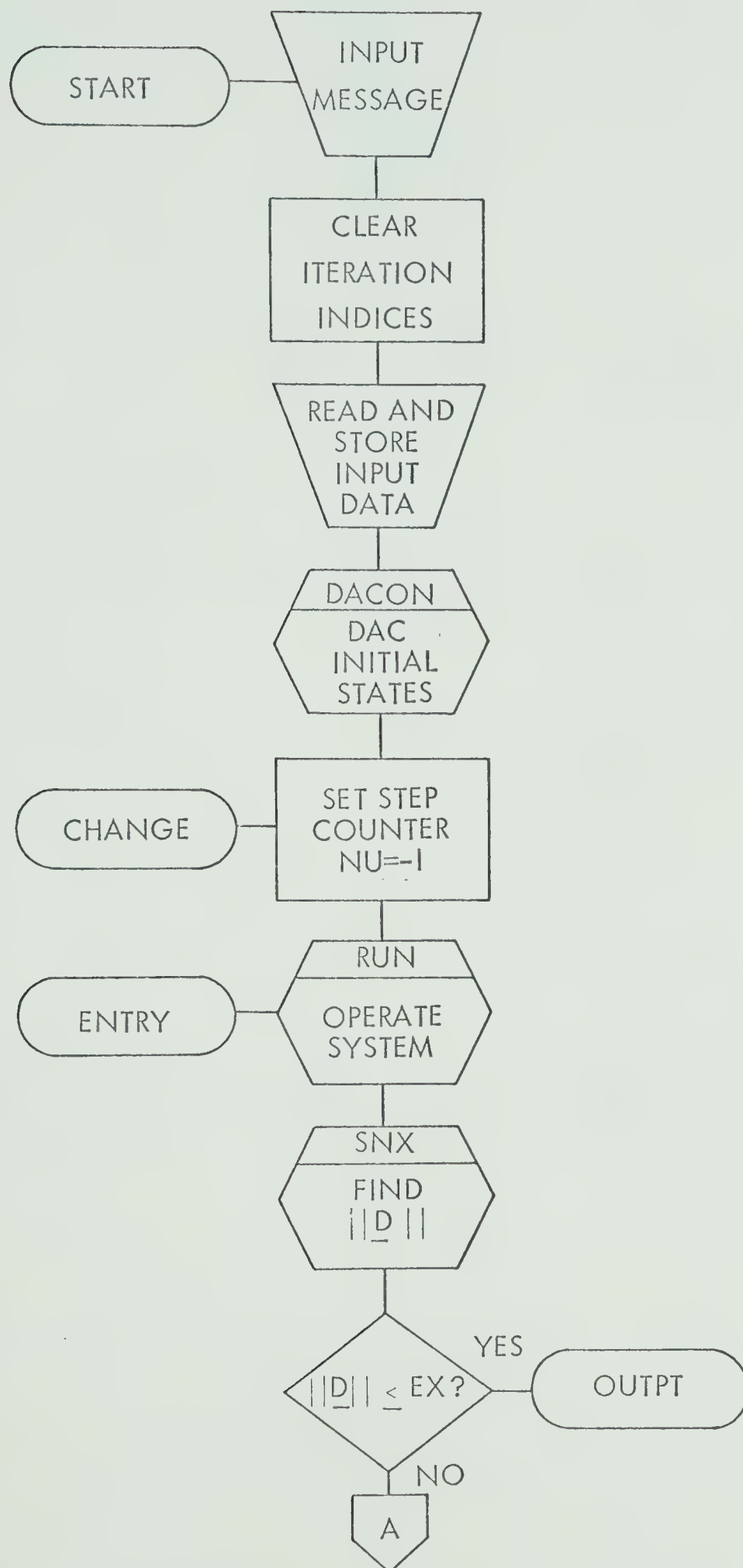
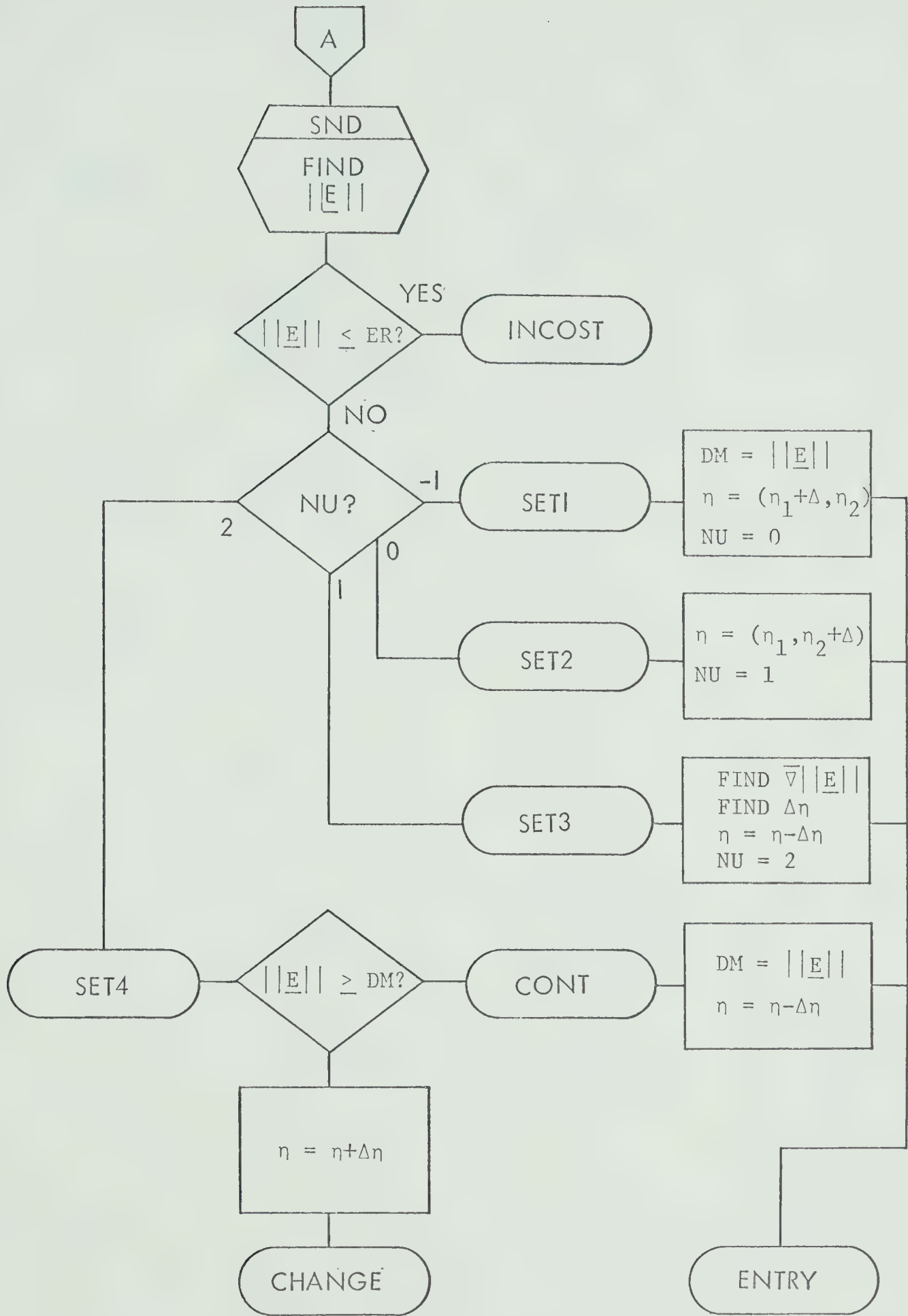
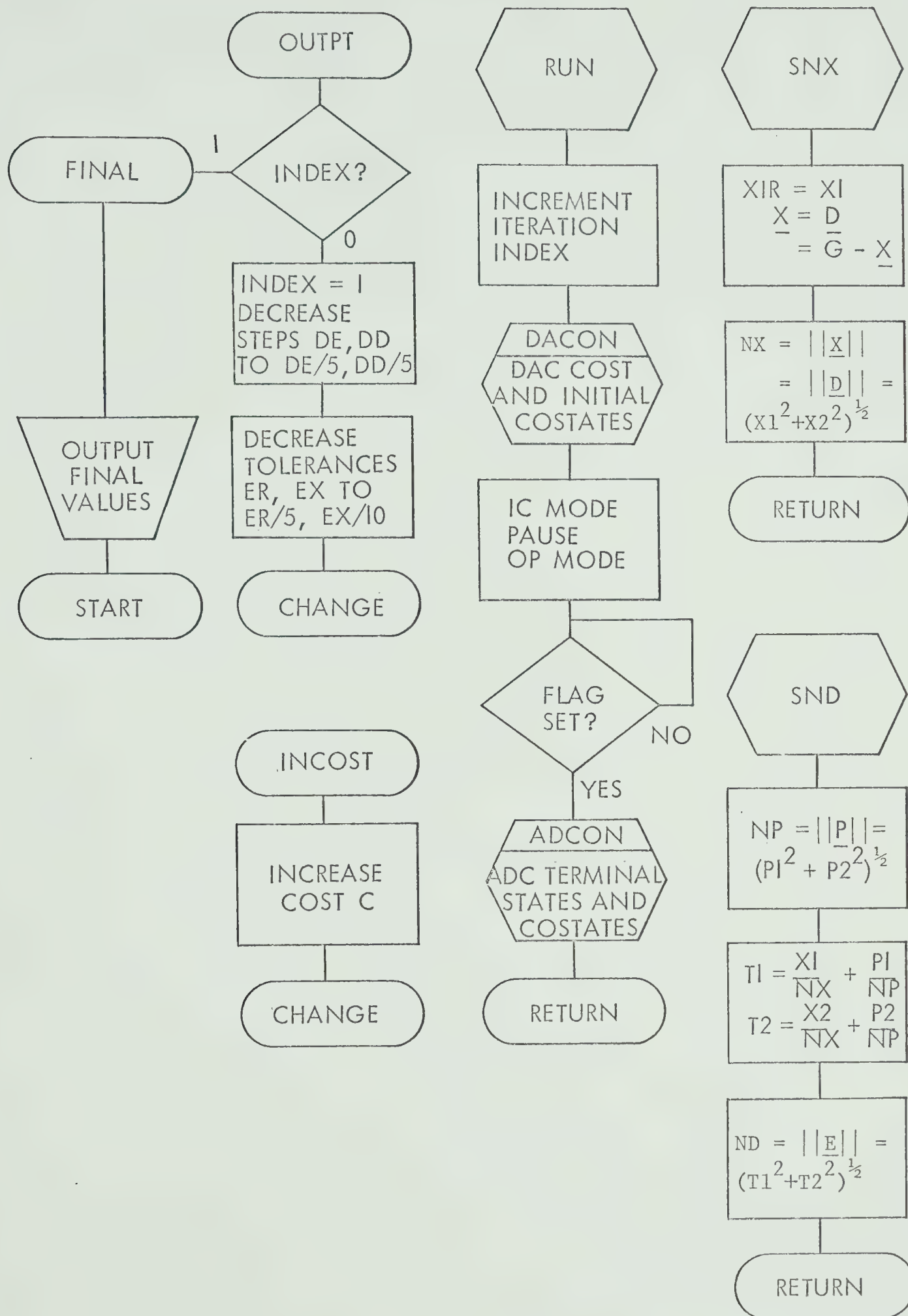


FIGURE A - 4.1





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